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Test of Spectral/Spatial Classifier

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16. Abstract This report provides the final results of the subtask to test the spectral/spatial classifier (ECHO). This document reports on: <ol style="list-style-type: none"> 1 the programming of the Nonsupervised ECHO algorithms, 2 tests of the effects of the input parameters on six performance measures, and 3 comparison of the Nonsupervised ECHO classifier with the Supervised ECHO classifier and the perpoint classifier. <p>The Nonsupervised ECHO classifier identifies objects without the benefit of class statistics. Statistics of the objects thus identified may be of value in training for the classifier. The Supervised ECHO classifier demonstrates superior classification accuracy, reduced variability of classification results, and requires less CPU time when compared to the perpoint classifier. The Nonsupervised ECHO processor requires less CPU time and produces less variable classification results than the perpoint classifier, but does not produce classification results which are superior to the perpoint classifier.</p>					
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TASK 2 TEST OF SPECTRAL/SPATIAL CLASSIFIER

INTRODUCTION

Contemporary classifiers for analysis of remotely sensed data compare spectral measurements from each feature of each pixel to class statistics, computing a likelihood discriminant function associated with each class, and categorizing the point according to the class with the largest discriminant function value. Each point is classified on the basis of its spectral measurements alone. One premise of this technique is that the objects of interest are large in comparison to the size of a point. If this were not so, a large proportion of points would be composites of several classes, making statistical pattern classification unreliable since pre-specified categories would be inadequate to describe actual states of nature. From this premise it follows that objects are represented by arrays of points, and that a statistical dependence exists between consecutive points. Contemporary classifiers fail to exploit the statistical dependence between adjacent points when assigning classes.

The ECHO processors benefit from spatial information by first aggregating into groups points whose spectral responses are not significantly different in a statistical sense, and then applying a maximum likelihood classification rule to these homogeneous groups. Homogeneous objects are identified in a three step process. First, cells are formed by systematically partitioning the data into N by N sized blocks of pixels. The statistics of each cell are then compared to a homogeneity criterion. Points which do not comprise homogeneous groups are classified on a point-by-point basis, just as contemporary classifiers classify all points. Statistics of adjoining homogeneous cells are then compared to annexation thresholds. Adjoining cells which appear to belong to the same statistical population are combined into a single object.

Two separate ECHO algorithms have been developed. The first, Supervised ECHO, makes use of pre-specified class statistics to identify homogeneous objects. The second, Nonsupervised ECHO, identifies homogeneous objects without the use of class statistics. Consequently, those objects identified by the Nonsupervised algorithm may be used to aid in the training process.

Past Work at LARS

Much of the background research on the ECHO concept was performed at LARS during 1975 and is documented in the Final Report for 1975[1], R. L. Kettig's doctoral thesis[2], a LARS Information Note[3], and in symposium proceedings[4].

*ECHO stands for Extraction and Classification of Homogeneous Objects.

The work related to the development, testing, and documentation of the Supervised ECHO algorithms is documented in Volume I of the Final Technical Report on NASA Contract NAS9-14970, June 1, 1976 - May 31, 1977 [5]. Included in that report are a more detailed background on the past development work on ECHO at LARS, an appendix containing program listings and documentation for the Supervised ECHO algorithms, the results of systematic tests of the Supervised processor on MSS data for agricultural regions as observed by the Landsat satellites, aircraft scanners, and simulated Thematic mapper data, and an example product of an object map enabling the determination of the utility of object maps to a LACIE Analyst Interpreter in the selection and labeling of training fields.

Objectives

The objectives for the contract extension were to:

- * Complete the Fortran implementation and documentation of the Non-supervised ECHO (Extraction and Classification of Homogeneous Objects), making the algorithms available for use by JSC.
- * Classify, at different parameter settings, Landsat, simulated Thematic Mapper and aircraft data sets. Compare the resulting CPU time required, full field performance, field center pixel performance and classification variability of results with those achieved by the perpoint and Supervised ECHO processors.
- * Evaluate the objects identified by the Nonsupervised algorithms in terms of two types of errors:
 1. more than one field on the ground being identified by ECHO as a single object, and
 2. a single field on the ground being subdivided into distinct objects.

DESCRIPTION OF WORK

Produce Documented Fortran Programs

Fortran listings and program abstracts for the Nonsupervised ECHO processor are presented in Appendix A. This processor performs field extraction without the benefit of class statistics. Statistics are necessary for classification of the objects identified, however. The software is designed to function in a two phase mode. The first phase is the Nonsupervised Field Extraction Algorithm. This phase of the processor partitions the data to be classified into a set of points and homogeneous fields and calculates the channel means and covariance matrix for each homogeneous field which is identified.

The Nonsupervised processor utilizes a homogeneity test which compares σ_{ij}/μ_{ij} to a threshold t_i ; where i refers to feature or channel and j refers to the cell. If the standard deviation divided by the mean for feature i exceeds the user-specified threshold t_i , the cell is "singular" and elements of the cell will be classified on a point-by-point basis. Adjoining homogeneous cells are annexed to fields on the basis of a two-step test, 1) that the channel variances of the field are equivalent to the channel variances of the cell and 2) that the channel means of the field are equivalent to the channel means of the cell. Should either of these criteria not be met for any channel, annexation will not take place.

The research software was designed to run in two phases. The first phase merely calculated the cell mean and covariance matrices and wrote them on tape; the second phase proceeded to perform field extraction followed by classification. This process has two disadvantages. First, although an object map could be produced, it had to be produced in phase two, the same phase that required a class statistics deck for input in order to perform the classification. Second, since only cell mean and covariance matrices were written on the intermediate tape, when singular cells were encountered in phase two, they had to be classified as small samples, a cell at a time, rather than on a point-by-point basis, because data values for individual points were not available to the phase two classifier.

The Nonsupervised ECHO software was restructured to move the field extraction algorithm into phase one of the processor and produce an intermediate tape containing:

- * class means and covariance matrices for each homogeneous object identified.
- * an object map containing the mean for channel i of object j in every pixel of object j and the original data values for those pixels belonging to singular cells.
- * a tag array identifying whether a pixel falls in a singular cell (and should be classified individually), or in a homogeneous object and hence should receive a class assignment based on the sample classification of the object to which it belongs).

The second phase of the restructured Nonsupervised processor reads the intermediate tape and the class statistics deck, performs a maximum likelihood point-by-point classification on points falling in singular cells, and a maximum likelihood sample classification of the homogeneous objects identified. Figure 2a-1 presents the general processing flow for phase one (field extraction) of the Nonsupervised processor. The general processing flow for phase two (classification) is presented in Figure 2a-2. Inputs to outputs from the Nonsupervised ECHO processor's two phases are described in Table 2a-1.

The formats of the disk and tape files which support the Nonsupervised ECHO processors are presented in Appendix B. User documentation is available in the forms of an ECHO User's Guide[6] and an Echo Case Study[7].

Figure 2a-1

2a-6

GENERAL FLOW OF
NONSUPERVISED ECHO

PHASE 1

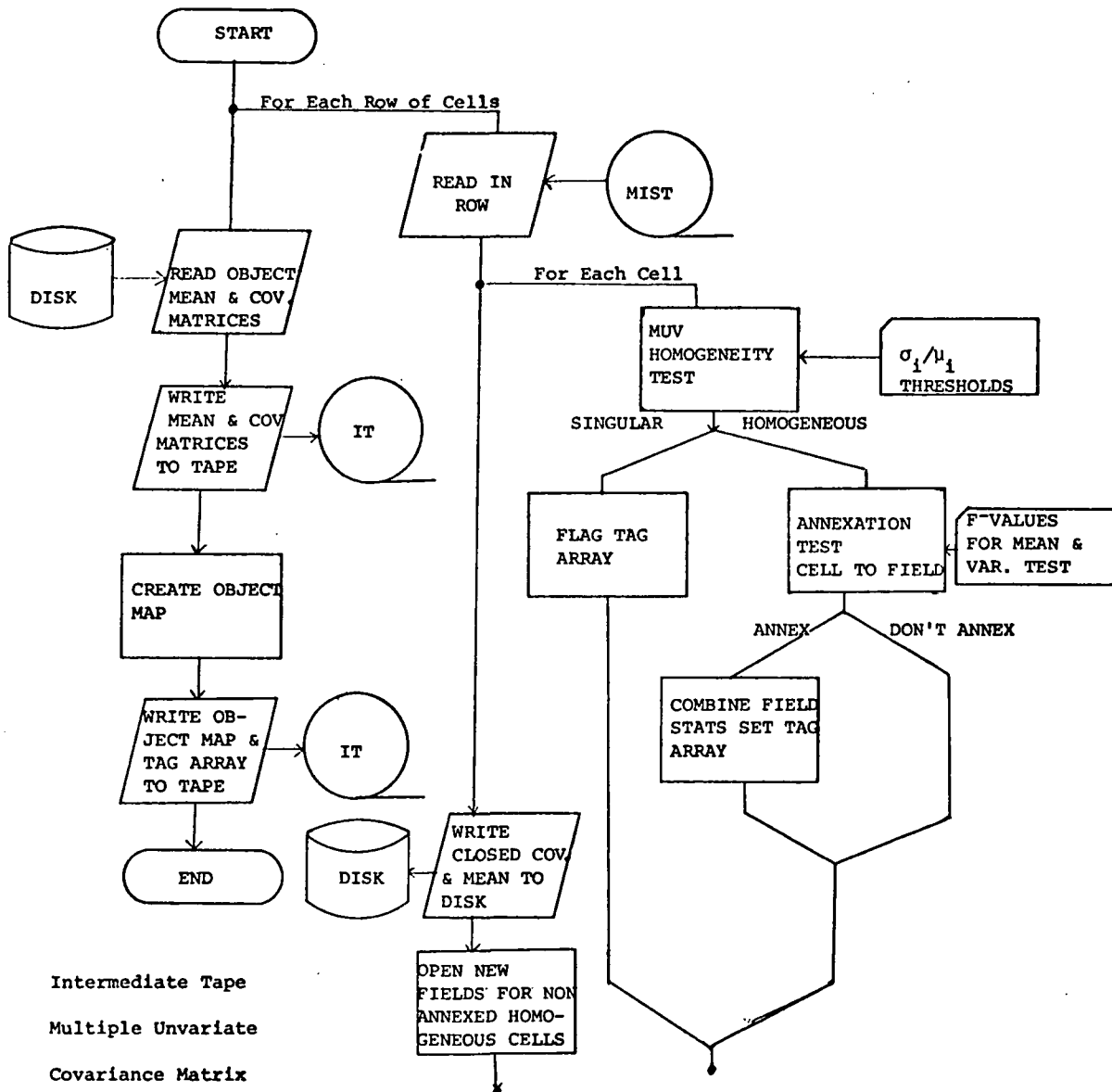


Figure 2a-2
GENERAL FLOW OF
NONSUPERVISED ECHO
PHASE 2

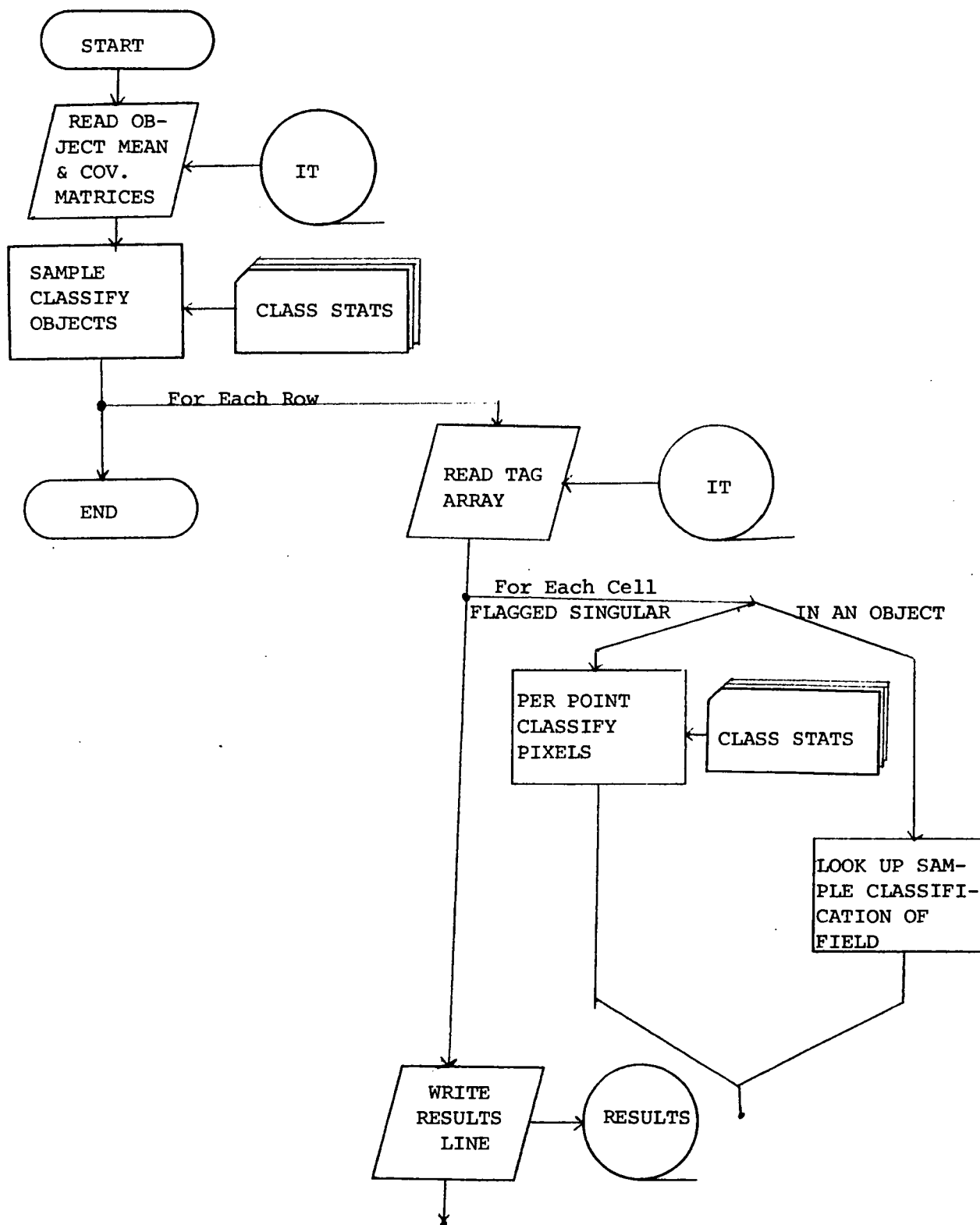


Table 2a-1
NONSUPERVISED ECHO
PROCESSOR

INPUTOUTPUTPhase One:

Channel Selection

Intermediate Tape containing

Cell Width

object map, object statistics

Cell Homogeneity Thresholds

and pixel tag array.

 $(\sigma_i/\mu_i \text{ for channel } i)$

Annexation Thresholds

(mean and variance)

Multispectral Image Storage

Tape

Intermediate Results Tape

and File Specification

Phase Two:

Class Statistics

Results File

Intermediate Tape from

Phase One

Specification of Results

Tape and File

Test the Nonsupervised ECHO Algorithms

1. Data Sets

The second objective of the ECHO Extension work is to test the Nonsupervised ECHO algorithms on MSS data for agricultural regions. Data sets are to include Landsat, aircraft, and simulated Thematic Mapper data. Nine Landsat, one aircraft and six simulated Thematic Mapper data sets (from resolutions from one site and two resolutions from a second site) were selected for analysis. The data sets are summarized in Table 2a .

Table 2a-2

Data Sets

(LANDSAT)

	CHANNEL NUMBER	WAVE BAND μ M
	1,5,9	.5-.6
	2,6,10	.6-.7
	3,7,11	.7-.8
	4,8,12	.8-1.1

Area	Channles Used	Data Collected	Data Set
Graham County, Kansas	9, 10, 11, 12	5/26/74	LACIE/SRS
Grant County, Kansas	5, 6, 7, 8	5/19/74	LACIE/SRS
Haskell County, Kansas	9, 10, 11, 12	5/27/74	LACIE/SRS
Kearny County, Kansas	9, 10, 11, 12	5/27/74	LACIE/SRS
Huntington County, Indiana	1, 2, 3, 4	7/7/73	CITARS
Shelby County, Indiana	1, 2, 3, 4	9/7/73	CITARS
White County, Indiana	1, 2, 3, 4	8/21/73	CITARS
Fayette County, Illinois	1, 2, 3, 4	8/21/73	CITARS
Lee County, Illinois	1, 2, 3, 4	8/5/73	CITARS

Table 2a-2 (Continued)

Simulated Thematic Mapper

TM CHANNEL NUMBER	WAVE BAND μM
1	.45-.52
2	.52-.60
3	.63-.69
4	.74-.80
5	.80-.91
6	1.55-1.75
7	10.4-12.5
8	.74-.91

Area	Channels Used	Date	Resolution
Williams County, ND	1, 2, 3, 6, 8	8/15/75	30m
Williams County, ND	1, 2, 3, 6, 8	8/15/75	40m
Finney County, Kansas	2, 3, 5, 6, 7	7/6/75	30m
Finney County, Kansas	2, 3, 5, 6, 7	7/6/75	40m
Finney County, Kansas	2, 3, 5, 6, 7	7/6/75	50m
Finney County, Kansas	2, 3, 5, 6, 7	7/6/75	60m

Table 2a-2 (Continued)

Area	Aircraft	Data Collected
Wavebands Used (μ M)		
Finney County, KS	.4-.49, .59-.64, .65-.69,	
	.82-.88, 1.53-1.62, 10.1-11.0	7/6/75

2. Training Procedures

a. Landsat Data Sets

The LACIE training sets were created using ground truth information provided by JSC for both test and training fields. Odd-numbered fields appearing in each subclass were used for training; even-numbered fields were used for test. For Graham and Grant counties, statistics were generated by using the STATISTICS processor. For Haskell and Kearny counties statistics were secured by clustering the training fields of each class into four subclasses which were then pooled on the basis of output from the SEPARABILITY processor. It should be noted that both training and test fields for the LACIE/SRS data are large enough to inset field boundaries approximately two pixels inside estimated field boundaries. This inset allows for any image misregistration which may occur between any two dates on the multitemporal input runs. Proportion estimate for the 1974 LACIE/SRS segments were available in ground truth packets provided by JSC.

The CITARS training sets were originally created by a supervised procedure using ground truth information provided by the Agricultural Stabilization and Conservation Services (ASCS) to select both the training and test fields^[8]. Those training sets were used without change in the ECHO tests.

Five counties in the CITARS experiment were used as test sites for the Nonsupervised processor evaluation. A data set free of clouds which occurred late in the growing season was required for each of the five counties. Dates from July, August or early September were selected. The classification results for all of the CITARS experiments are catalogued on a series of LARS tapes. The desired data sets were located on the catalogued CITARS tapes and the statistics which had been used for the CITARS experiment were obtained by using the LARSYS PUNCHSTATISTICS processor. The pooling of classes was determined by running the PRINTRESULTS processor and requesting training field results. This combination of requests produced a table of available classes in the statistics deck and also the informational names under which they were classified. By using this list it was possible to reconstruct the combinations of spectral classes and pooled spectral classes which were needed to reproduce the original classification.

Training and test fields for each CITARS classification were secured and appropriate control cards added to evaluate the ECHO classifications of CITARS data sets. The proportion estimates used for the CITARS evaluation are estimates of the proportions of the various classes for the entire county made by the SRS.

b. Simulated Thematic Mapper Data Sets

Training sets from the simulated Thematic Mapper tests performed at LARS in 1976 were used for ECHO analysis. These training sets were generated by selecting fields of known cover types and clustering each informational class separately to define subclasses.

Color infrared photographic mosaic prints were made from photographic data collected concurrently with the scanner data. Informational class information provided by ground observations was transferred to clear plastic overlays on the mosaic print. The analyst could then easily locate the corresponding fields in his cluster maps and assign the field coordinates to the informational classes.

Statistics were calculated for each training area and compared using the SEPARABILITY processor. Similar classes were combined, where indicated and the data set was used to classify the flightline. Training areas were not excluded from the test fields since the test fields had been pre-selected for the entire flightline.

Two to four subclasses were found in each informational class. The Kansas flight was an exception. Because of severe line-to-line changes in signal level in the original 6 meter scanner data, the analyst was forced to create more spectral classes to account for the within-class variations due to excessive noise. This was most apparent in the 30 and 40 meter resolution data. The effect was reduced but not eliminated in the 50 and 60 meter resolution data. Alternate fields were used for training and test decks, respectively. The procedure was repeated for each of the four resolution sizes. As resolution size increases, the number of spectral classes decreases.

The entire training set selection procedure was repeated for each resolution size so that any effects on training set selection which might be caused by data resolution would be included in the analysis results. An example is the increasing difficulty and eventual impossibility of selecting samples from small, or narrow, fields as the resolution size increases.

c. Aircraft

A 6-meter aircraft scanner data set used to generate the Finney County, Kansas simulated Thematic Mapper data set was used. The same training and test fields used for the Thematic Mapper were available in the six meter data set. Because of the very large number of data points in the full data set, only the first two miles (one third of the total flight line) were used for the Nonsupervised ECHO evaluation.

The data set was not corrected for sun and scanner angle effects. To compensate for these angle effects, training fields were distributed across the width of the flight line. At least six fields in each informational class were used in the training set. Fields from informational classes were clustered together into five spectral classes. All spectral classes from all informational classes were combined into a statistics deck and appropriate pooling was done based on SEPARABILITY results. The classification results were strongly influenced by angle effects.

3. Dependent Variables

There are six variables which were monitored to evaluate the ECHO algorithm.

- CPU time
- Field center pixel classification performance
- Training field classification performance
- Full field classification performance
- RMS proportion estimate error
- Classification variability

These variables are related to reasons for adopting a new classification technique: cost, accuracy, and usability of results. The CPU time required to perform a classification is one way to measure the cost of classification. Field center pixel, full field, and training field performances and RMS proportion estimate error are all ways to evaluate the accuracy of the classifier. Classification variability is a measure of "salt and pepper effect" in classification results.

The CPU time required to execute each of the ECHO classifications has been recorded so that the effects of varying the cell homogeneity and annexation thresholds may be monitored. The CPU time required to perform the perpoint classifications have been adjusted to reflect the increased efficiency of the LARSYS perpoint classifier which is coded in assembly language. Thus, the CPU time recorded for a perpoint classification is what a FORTRAN classifier would have required to perform the classification.

The indices of classification performance were applied in several ways. Classification accuracy (identification) was evaluated utilizing field center pixel, "full field" and test field sample performances for all data sets. Proportion estimation was carried out for the Landsat and Simulated Thematic Mapper data sets.

The training performance is the overall classification accuracy (number of training pixels correctly classified divided by the total number of training pixels) of the pixels used to calculate the class statistics. Field center pixel performance is the overall classification accuracy of pixels inset at least one pixel from the field boundary. For the registered LACIE/SRS data the field center pixels are inset at least two pixels from the field boundary. Although this procedure insures that the pixels examined are not mixture pixels, it has the unfortunate effect of eliminating smaller fields from consideration. The third measure of classification accuracy, "full field" performance, includes those pixels on the boundaries of the fields in the classification performance. The "full field" pixels were generated by expanding the field center pixel boundaries one pixel in all directions.

The RMS error of informational class proportion estimates for each flightline was found by calculating the percent of the flightline classified as a particular class and comparing it with the ground-collected estimate using equation (1).

$$\text{RMS Error} = \sqrt{\frac{\sum_{i=1}^N (C_i - C'_i)^2}{N}} \quad (1)$$

where, N = number of informational classes,
 C_i = percent classified as informational class i, and
 C'_i = percent of class i estimated from ground-collected data.

RMS error is calculated for the Landsat and Thematic Mapper data runs. The Agricultural Stabilization and Conservation Service (ASCS) provided the ground truth proportion estimates for the simulated Thematic Mapper data set. Proportion estimates for the 1974 LACIE/SRS segments were provided in ground truth packets received from JSC. The SRS county proportion estimates were used to calculate RMS proportion error for the CITARS data set.

Average variability is a measure of the rate of change from one information class to another. It should reflect the degree to which ECHO reduces the "salt and pepper effect" which is sometimes present in per-point classifications. Variability is calculated by systematically selecting 50 lines of the classified area, counting the number of information class changes, and dividing by the number of opportunities for class changes.

$$\text{Variability} = \text{NCC} / (50 * (\text{NS} - 1)) \quad (2)$$

Where:

NCC = the number of class changes over the 50 selected lines, and
 NS = the number of classified pixels/lines.

4. Results

This section outlines the results of the tests performed on the Nonsupervised ECHO classifications and the comparison of the Nonsupervised ECHO processor with the Supervised ECHO and the perpoint classifiers. Results are discussed separately for each scanner type, Landsat, simulated Thematic Mapper, and aircraft.

For all data sets considered, the Nonsupervised ECHO processor was run with the following parameter settings:

1. Cell width of 2.
2. Cell homogeneity thresholds of 0.05, 0.10, and 0.25.
3. Both the mean and the covariance annexation thresholds set at 0.001, 0.01, and 0.1.

a. Landsat Results

Training and test information for the nine Landsat test data sets were drawn from the 1974 Kansas LACIE/SRS data sets and the 1973 CITARS data sets. Four LACIE/SRS and five CITARS data sets were considered.

1 LACIE/SRS Results

Figures 2a-3 through 2a-8 present the average results for the four LACIE/SRS sites examined at a cell width of two for three homogeneity and three annexation parameters. Figure 2a-3 plots the average CPU time in seconds required by the perpoint classifier (represented by the line of 'P's) versus the average CPU time required by the Nonsupervised ECHO routine to classify the four LACIE/SRS data sets at each of nine Nonsupervised ECHO parameter settings for 2 by 2 pixel cells. The cell homogeneity threshold is plotted along the horizontal axis. As this threshold increases, it becomes more likely a cell will be classified as a unit, less likely that a cell will be split and its constituent pixels classified individually. The dependent variable, CPU time, appears on the vertical axis. The cell-to-field annexation parameter for each cell homogeneity threshold is represented on the plot. A '1' appears in the position for the results achieved when both the mean and covariance annexation thresholds are set to 10^{-1} ; a '2' when they are set to 10^{-2} ; and a '3' when they are set to 10^{-3} . As the annexation thresholds become smaller, the likelihood that adjoining homogeneous cells will be annexed into a single field increases. When two or more annexation thresholds achieve the same performance, a star appears in the position on the plot.

The statistical significance of the effects of the Nonsupervised ECHO parameters on the LACIE/SRS results are presented in Table 2a-3.

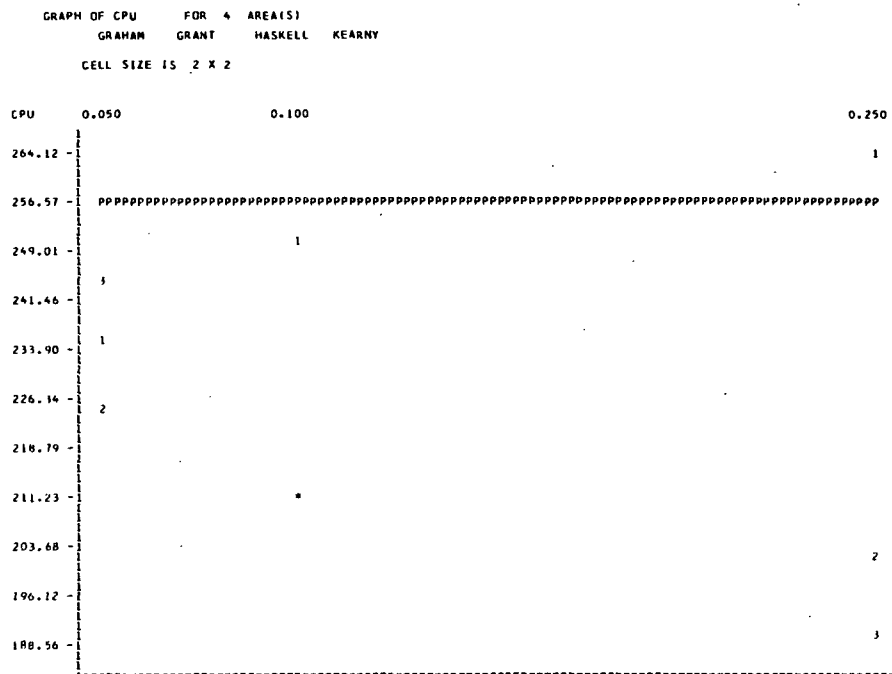


Figure 2a-3
Nonsupervised ECHO CPU Requirements
for the LACIE/SRS Data Sets

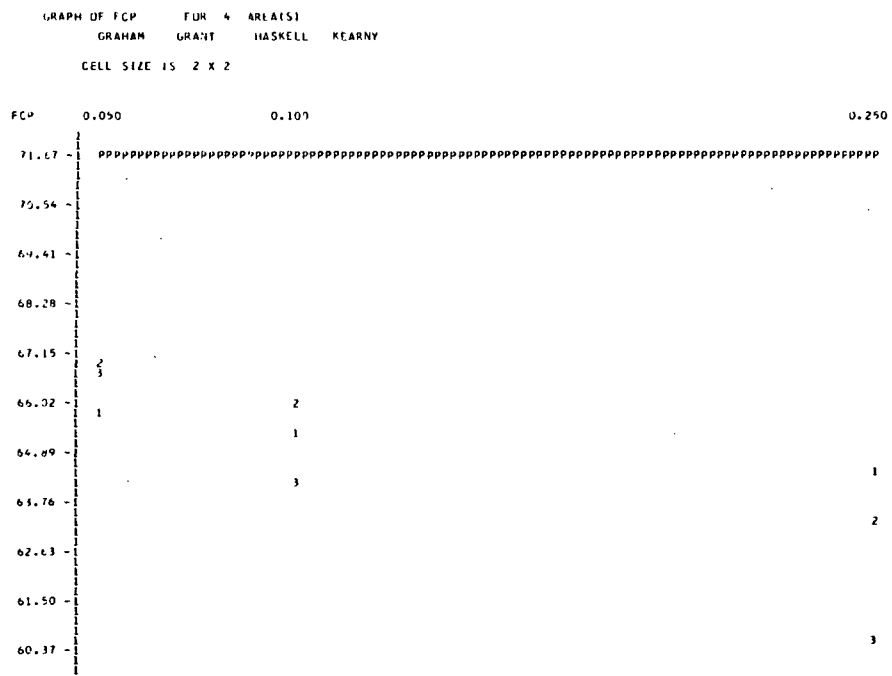


Figure 2a-4
Nonsupervised ECHO Field Center Pixel Performance
for the LACIE/SRS Data Sets

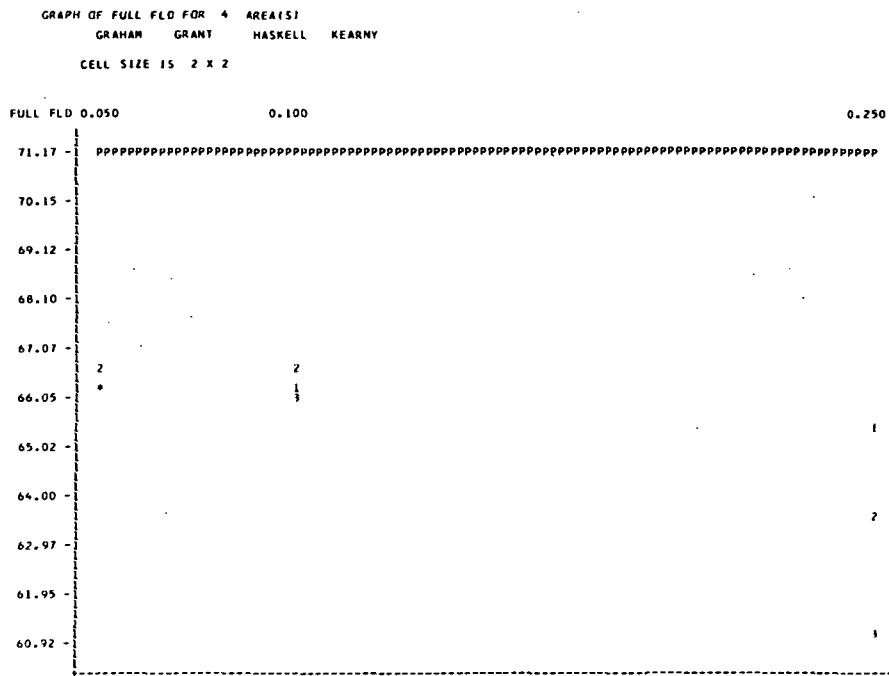


Figure 2a-5

Nonsupervised ECHO Full Field Performance
 for the LACIE/SRS Data Sets

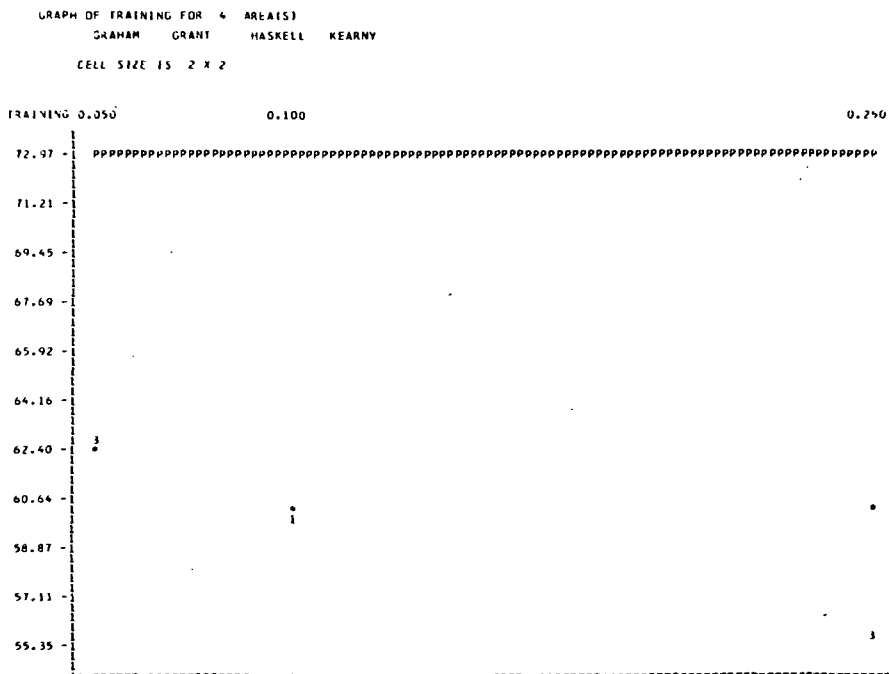


Figure 2a-6

Nonsupervised ECHO Training Field Performance
 for the LACIE/SRS Data Sets

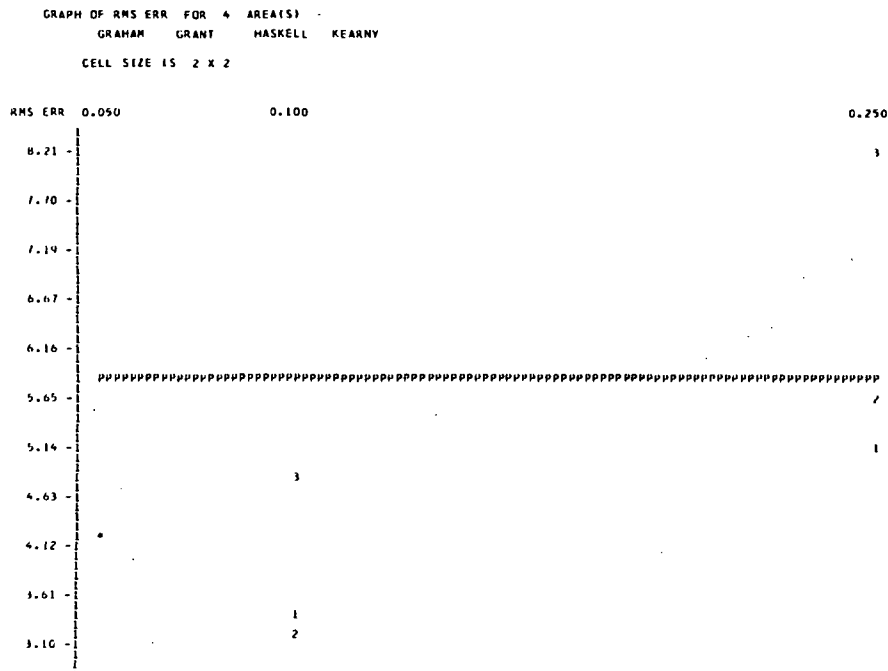


Figure 2a-7

Nonsupervised ECHO RMS Proportion Estimate Error
for the LACIE/SRS Data Sets

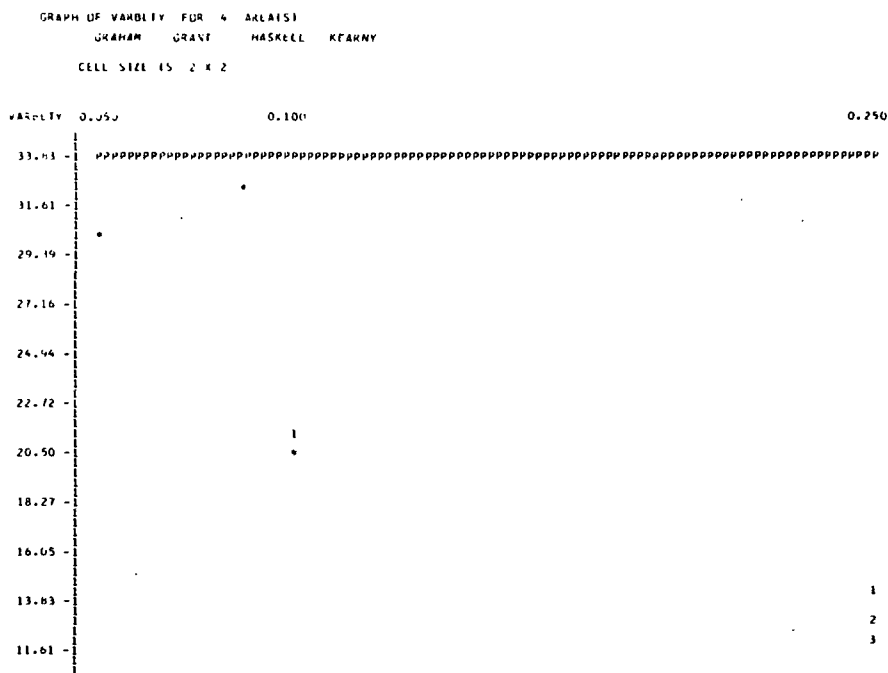


Figure 2a-8

Nonsupervised ECHO Classification Variability
for the LACIE/SRS Data Sets

Table 2a-3
Overall Landsat Comparisons
Effects of Parameters on LACIE Data

<u>Variable</u>	<u>Homogeneity Thresholds</u>	<u>Annexation Thresholds</u>	<u>Homogeneity X Annexation</u>
Degrees of Freedom	2,6	2,6	4,12
CPU Time	1.86	6.69*	21.37***
Field Center Pixel Performance	1.27	7.02*	1.90
Full Field Performance	1.31	11.09**	3.74*
Training Field Performance	.70	3.60 ⁺	5.89**
RMS Proportion Error	.86	4.03 ⁺	1.18
Classification Variability	13.64**	14.39**	12.59***

Significance Levels

Table entries are F-values

+	10%
*	5%
**	1%
***	.1%

It can be seen from Figure 2a-3 that, as the cell homogeneity parameter increases from .05 to .25, the CPU time required is effected by the interaction between the homogeneity and the annexation threshold. This interaction is probably the result of two separate influences. The LACIE/SRS, CITARS, and aircraft data runs of the Nonsupervised processor were produced utilizing the "MAP" option. When this option is specified, the data value for each point which falls within a field is replaced by the mean value for the field. This replacement requires a significant amount of computer time. Therefore, as more fields are identified, more CPU time will be used. As the homogeneity parameter increases, there is a tendency for more cells to be identified as homogeneous, and therefore, it is possible for more fields to be identified. The effect of this is reflected by the increase in CPU time of the runs with a 0.1 annexation parameter. However, with annexation parameters of 0.01 and 0.001, adjoining homogeneous cells are frequently annexed into fields reducing the number of passes through the classification equation and reducing the amount of computer time required.

Figure 2a-4 graphs the field center pixel performance for the Nonsupervised ECHO classifier. The only significant effect is that as the annexation parameter goes from 10^{-1} to 10^{-3} , the field center pixel performance declines.

For training field and full field performance measures presented in Figures 2a-5 and 2a-6, as the annexation threshold becomes smaller the performance decreases and as the homogeneity value becomes larger the effects of the annexation parameters are increased.

Parameter selection seems to have an only slightly significant effect on the RMS proportion estimate error (Figure 2a-7). As the annexation parameter becomes smaller, the RMS error tends to become larger.

As Figure 2a-8 indicates, classification variability is strongly influenced by both the annexation and the homogeneity parameters and by their interaction.

ii CITARS Results

Figures 2a-9 through 2a-14 present plots for the 2 by 2 cell size Nonsupervised ECHO results achieved over the five CITARS data sets. Characteristic differences between the CITARS and the LACIE/SRS data sets include:

- * The CITARS data set has a much smaller field size than the LACIE/SRS set.
- * The information classes are different. CITARS information classes are corn, soybeans, and other; LACIE/SRS classes are wheat and other.
- * the ground truth proportion estimates for the LACIE/SRS sites were for the area of the LACIE segment. The ground truth proportion estimates for the CITARS sites are for the whole county in which the data set lies, not for the area of the county which was actually sampled. Analysis of various results for the five CITARS data sets are presented in Table 2a-4.

Figure 2a-9 indicates that the CITARS data has the same interaction of annexation and homogeneity parameters as the LACIE/SRS data sets (Figure 2a-3) had with respect to CPU time required. This interaction is due to:

- * the additional CPU time required to replace pixel values with field means as additional fields are identified, and
- * the reduction of CPU time required by the classifier as larger fields are identified, resulting in fewer passes through the classification equation.

As the annexation thresholds go from 10^{-1} to 10^{-3} , the CPU time required to perform the Nonsupervised ECHO classification is reduced. For the CITARS data, as the homogeneity parameter increases, the decrease in average CPU time required is statistically significant.

There are no statistically significant parameter effects on the CITARS field center pixel or full field performance measures. Training field performance decreases as the annexation thresholds are decreased from 10^{-1} to 10^{-3} (Figure 2a-12). As the homogeneity parameter increases, the effect of the annexation parameters is increased for training field performance, RMS proportion error and classification variability measures.

As the homogeneity parameter increases, RMS proportion estimate error increases (Figure 2a-13) and classification variability decreases (Figure 2a-14). As the annexation thresholds decrease, the training field performance decreases, the RMS proportion error increases, and the classification variability decreases.

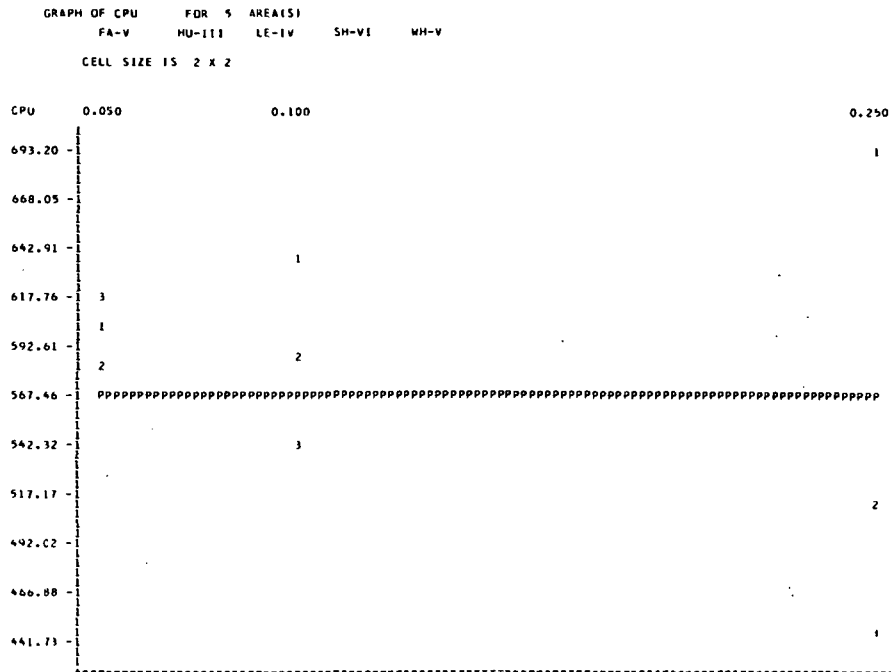


Figure 2a-9

Nonsupervised ECHO CPU Requirements for the CITARS Data Sets

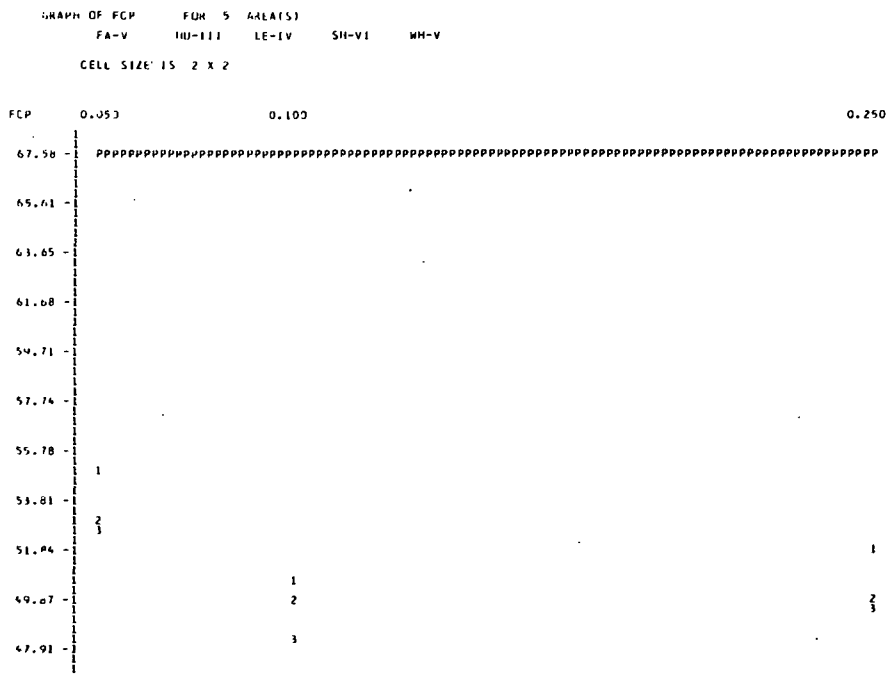


Figure 2a-10

Nonsupervised ECHO Field Center Pixel Performance for the CITARS Data Sets

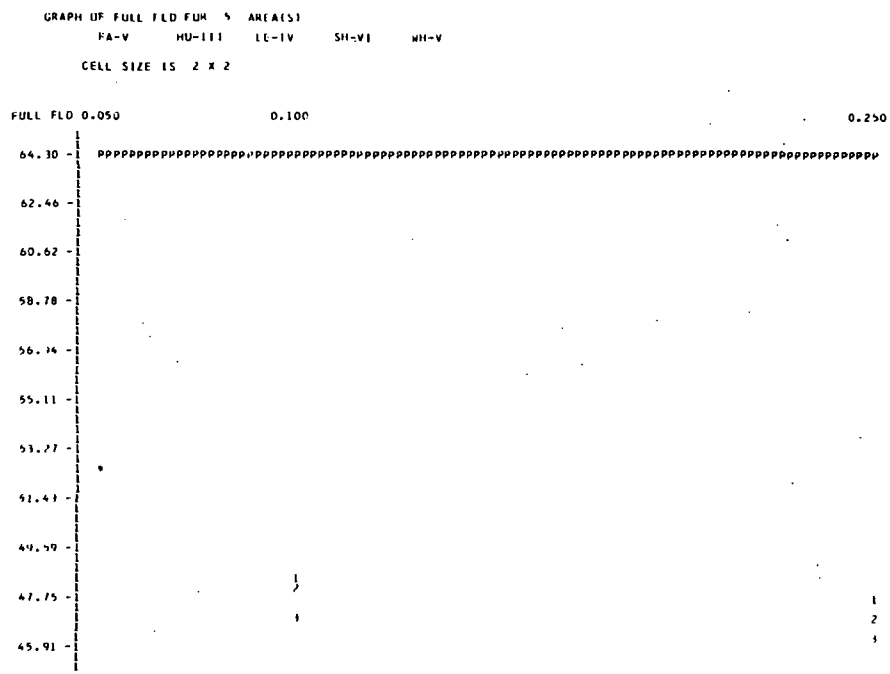


Figure 2a-11

Nonsupervised ECHO Full Field Performance for the CITARS Data Sets

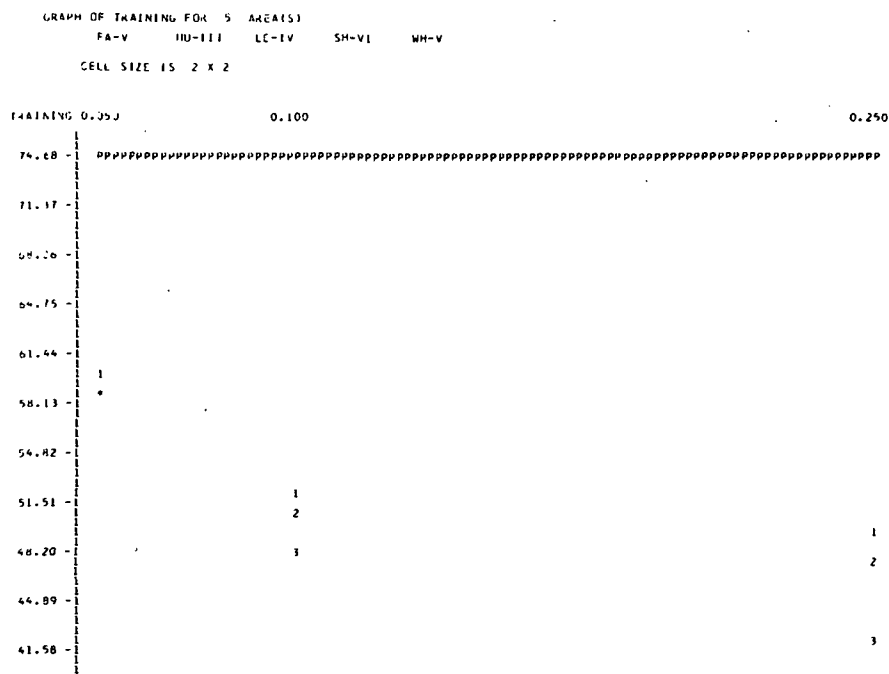


Figure 2a-12

Nonsupervised ECHO Training Field Performance for the CITARS Data Sets

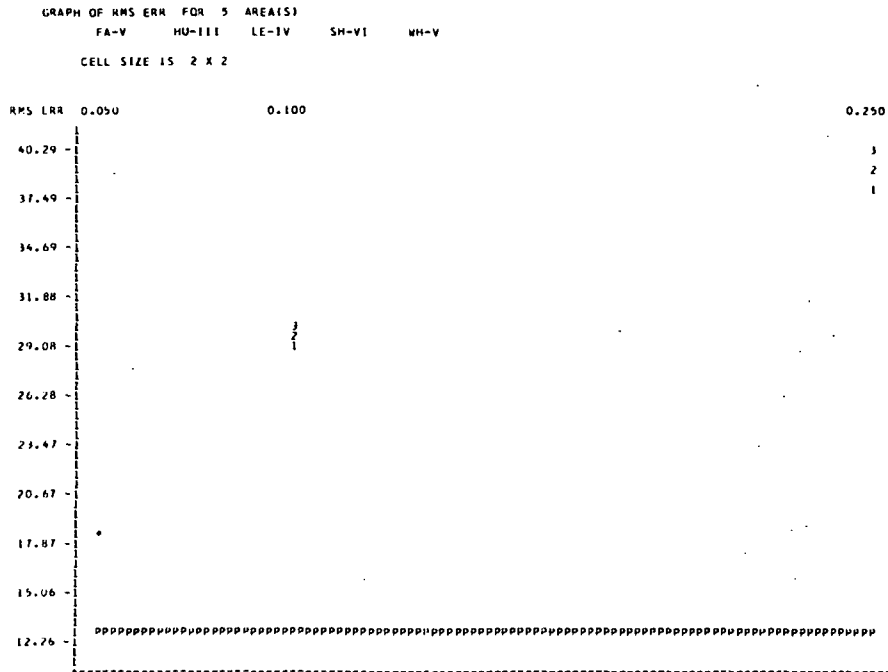


Figure 2a-13

Nonsupervised ECHO RMS Proportion Estimate Error
for the CITARS Data Sets

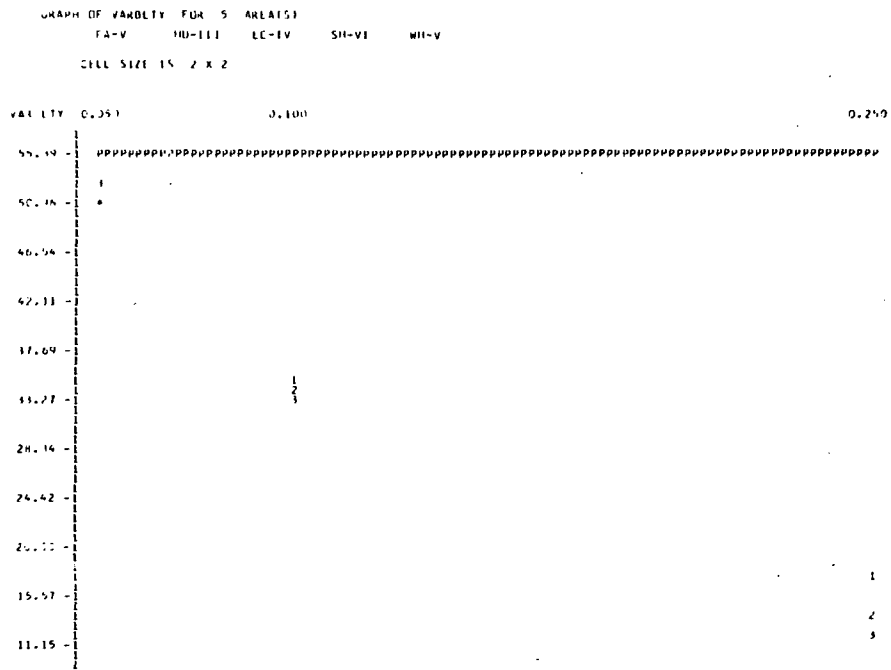


Figure 2a-14

Nonsupervised ECHO Classification Variability
for the CITARS Data Sets

Table 2a-4

Overall Landsat Comparisons
Effects of Parameters on CITARS Data

<u>Variable</u>	<u>Homogeneity Thresholds</u>	<u>Annexation Thresholds</u>	<u>Homogeneity X Annexation</u>
Degrees of Freedom	2,8	2,8	4,16
CPU Time	7.57*	13.73**	31.73***
Field Center Pixel Performance	.18	1.48	.16
Full Field Performance	.63	.84	.31
Training Field Performance	4.18 ⁺	8.62*	3.00*
RMS Proportion Error	10.29**	4.81*	4.22*
Classification Variability	158.47**	3.74 ⁺	7.81***

Significance Levels

+	10%
*	5%
**	1%
***	.1%

Table entries are F-values

In both the CITARS and the LACIE/SRS data sets, the classification performance measurements are markedly inferior to the perpoint performance over the same areas. This circumstance probably results from errors in object identification. It is of interest that while Nonsupervised ECHO performance measures for the LACIE/SRS data sets are 4 to 10 percentage points lower than the perpoint results, the average performance measures for the CITARS data sets are 10 to 30 percentage points inferior. This difference may be due to the effects of the smaller average field size for the CITARS data. The RMS proportion error for the Nonsupervised algorithm's results is superior to the perpoint algorithm results in the LACIE/SRS data, but inferior in CITARS case.

The Nonsupervised ECHO field center pixel performances are superior to the perpoint results in three of the five data sets (LE-IV, Hu-III, and Fa-V) by a few percentage points, but on the order of 17 to 20 points worse in the other two data sets (Sh-V and Wh-V). This indicates that the Nonsupervised processor has the potential to improve the classification accuracy achieved, but that it also may substantially decrease the accuracy.

iii Overall Landsat Results

Figures 2a-15 through 2a-20 summarize the effects of the ECHO parameter settings on the six dependent variables for nine Landsat data sets. Table 2a-5 summarizes the statistical significance of the effects of the ECHO parameter settings on these six dependent variables.

The homogeneity parameter has statistically significant effects on CPU time, training field performance, RMS proportion error, and classification variability. As the cell homogeneity threshold is increased, the amount of CPU time required to perform a classification decreases (Figure 2a-15). As the homogeneity parameter increases:

- * the training field performance increases (Figure 2a-18),
- * RMS proportion estimate error (Figure 2a-19) increases, and
- * classification variability (Figure 2a-20) decreases.

Parameter settings of the annexation thresholds have a significant effect on five dependent variables, CPU time, full field performance, training field performance, RMS proportion error, and classification variability. As the annexation threshold goes from 10^{-1} to 10^{-3} (the tendency for cell-to-cell annexation increases), the amount of CPU time, the full field performance, the training field performance, and the classification variability all decrease. As the annexation thresholds decrease, the RMS proportion estimate error increases.

The interaction effects of the homogeneity and annexation parameters were significant for four of the dependent variables, CPU time, training field performance, RMS proportion error, and classification variability. For these four variables, as the cell homogeneity parameter was increased, the effects of the annexation parameters were increased. This result is expected since annexation may take place only when adjoining cells are homogeneous.

Table 2a-6 presents the significant effects of the homogeneity and annexation parameters on the six dependent variables for each of the nine individual Landsat data sets. Although the effects of the homogeneity threshold on the field center pixel and full field performances are not significant when the data sets are considered together, the effect of the homogeneity parameter is significant in seven and eight of the nine Landsat data sets, respectively, when the data sets are analyzed individually. This result indicates that the effects of the homogeneity parameter on the field center pixel and full field performances are opposite for differing data sets.

On the other hand, the effects of the homogeneity and the annexation thresholds on CPU time are not statistically significant for any Landsat data set, when the data sets are examined individually.

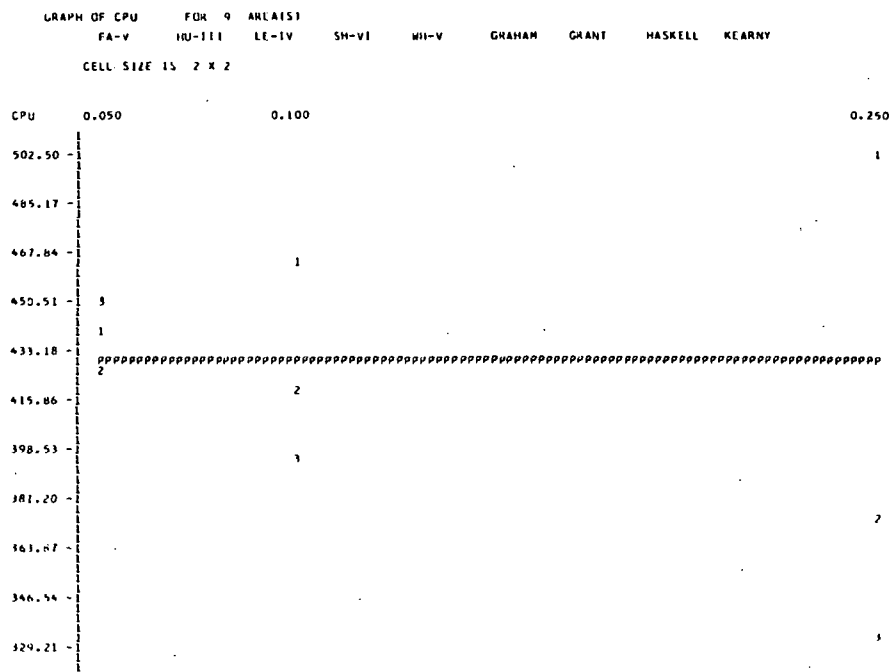


Figure 2a-15

Nonsupervised ECHO CPU Requirements
for all Landsat Data Sets

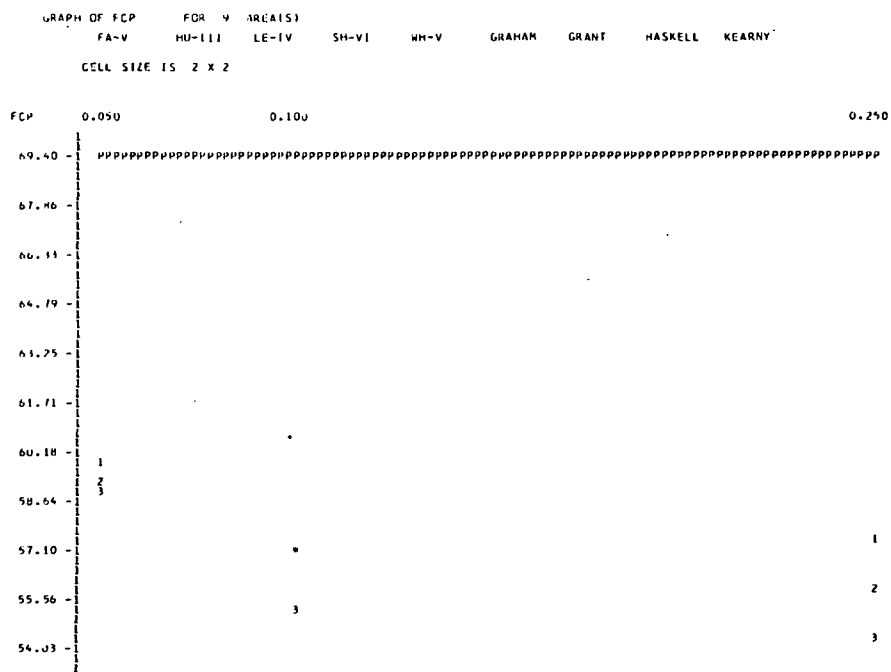


Figure 2a-16

Nonsupervised ECHO Field Center Pixel Performance
for all Landsat Data Sets

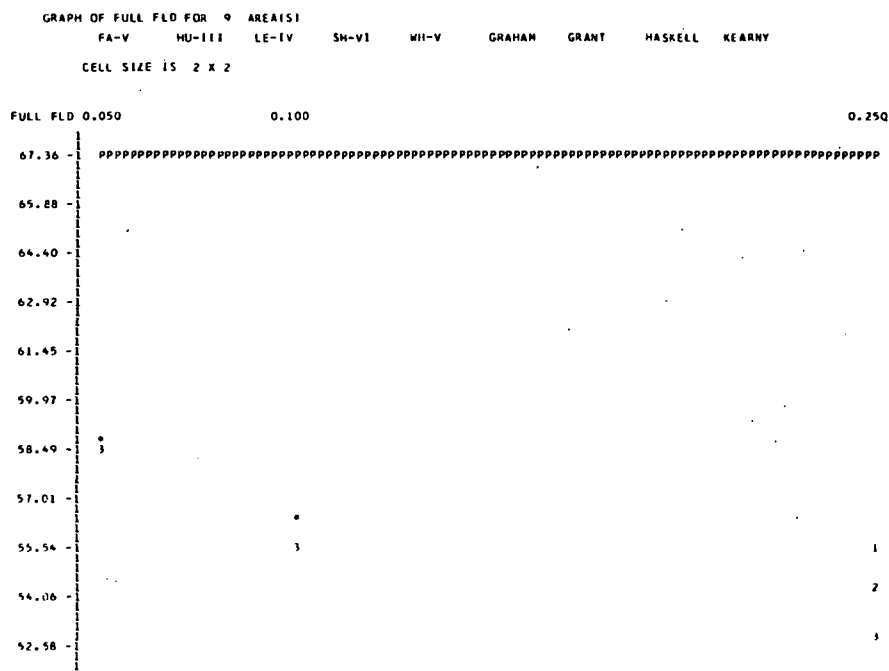


Figure 2a-17

Nonsupervised ECHO Full Field Performance
 for all Landsat Data Sets

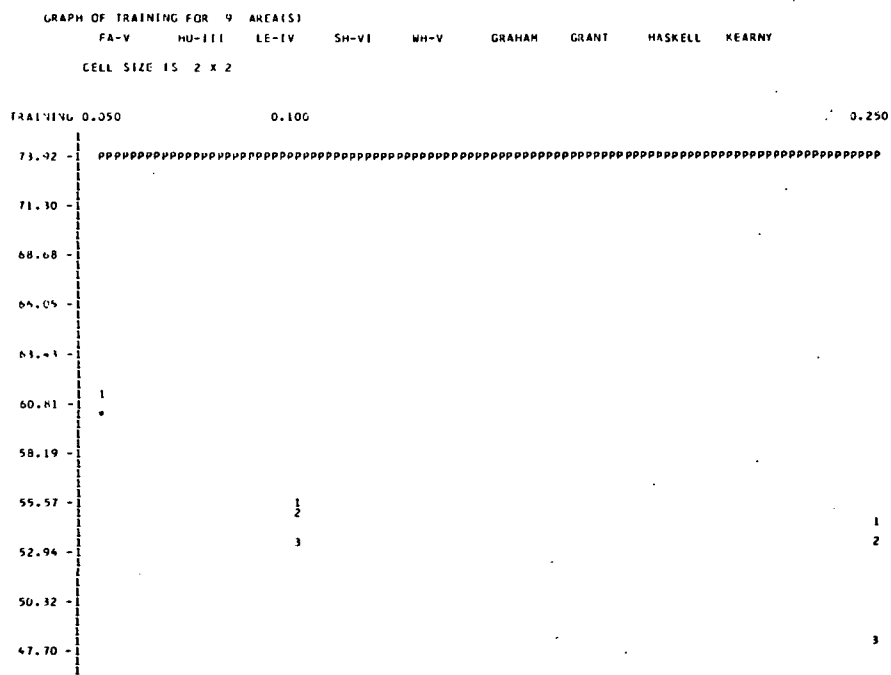


Figure 2a-18

Nonsupervised ECHO Training Field Performance
 for all Landsat Data Sets

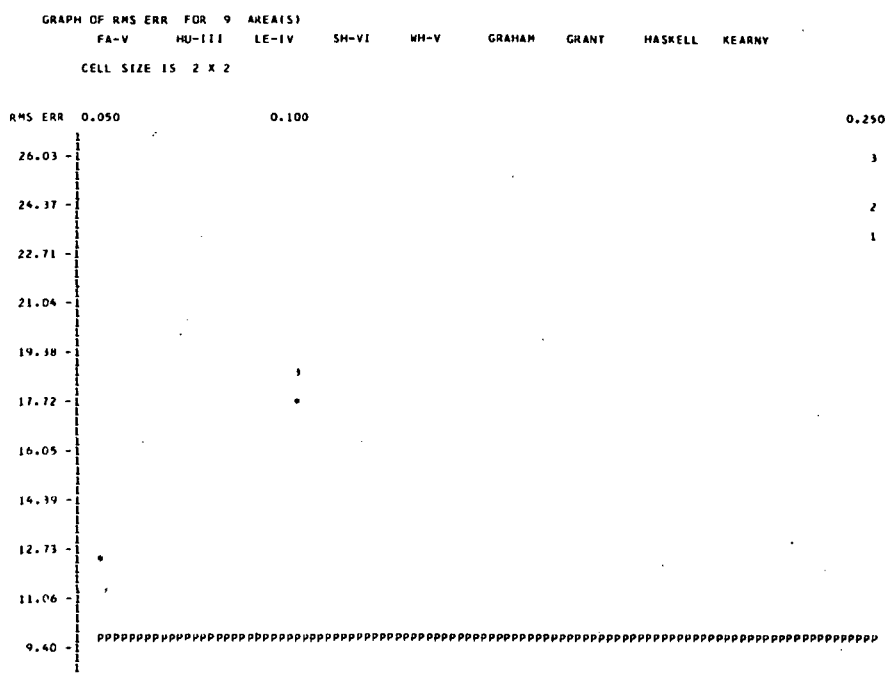


Figure 2a-19

Nonsupervised ECHO RMS Proportion Estimate Error
for all Landsat Data Sets

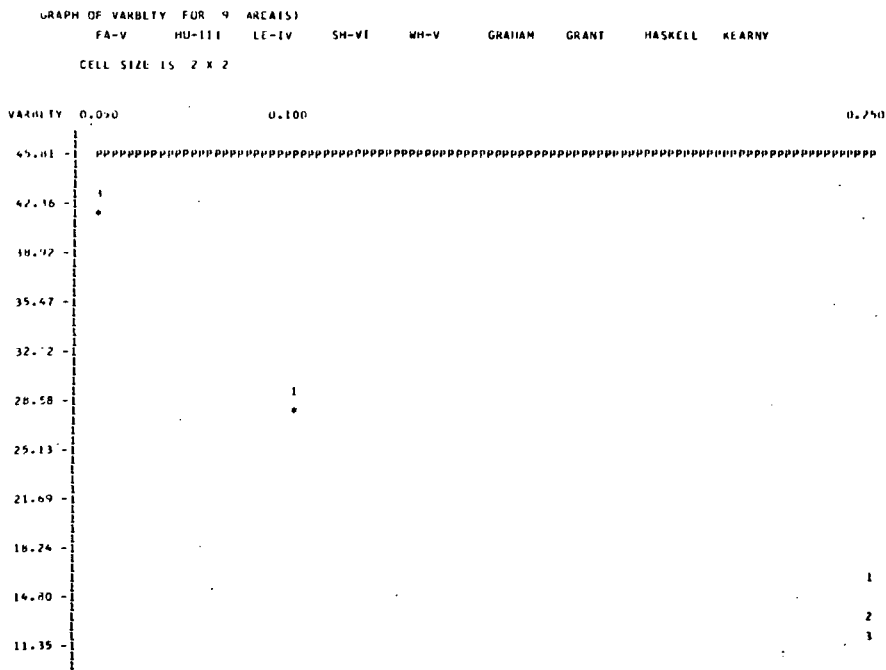


Figure 2a-20

Nonsupervised ECHO Classification Variability
for all Landsat Data Sets

TABLE 2a-5
Overall Landsat Comparisons
Effects of Parameters

<u>Variable</u>	<u>Homogeneity Thresholds</u>	<u>Annexation Thresholds</u>	<u>Homogeneity X Annexation</u>
Degrees of Freedom	2,16	2,16	4,32
CPU Time	7.61**	13.89***	19.08***
Field Center Pixel Performance	.45	2.92 ⁺	.90
Full Field Performance	1.11	3.89*	2.06
Training Field Performance	4.56*	9.46**	7.54***
RMS Proportion Error	6.48**	8.26**	3.68**
Classification Variability	43.08***	8.87**	8.96***

Significance Levels

+	10%
*	5%
**	1%
***	.1%

Table entries are F-values

Table 2a-6

Results of Individual Landsat Data Set Parameter Anovas
(F Tests with 2,4 degrees of Freedom)

	CPU	FCP	FULL	TRAIN	RMS	VARI
GRAHAM						
Sell	ns	ns	10%	ns	ns	0.1%
Ann1	10%	ns	ns	ns	25%	ns
GRANT						
Sell	ns	5%	5%	ns	0.1%	0.1%
Ann1	25%	10%	10%	ns	ns	5%
HASKELL						
Sell	ns	25%	25%	ns	25%	0.1%
Ann1	10%	ns	ns	ns	ns	10%
KEARNY						
Sell	ns	0.1%	0.1%	0.1%	1%	0.1%
Ann1	25%	ns	ns	25%	ns	25%
FA-V						
Sell	ns	1%	1%	ns	0.1%	0.1%
Ann1	ns	5%	10%	25%	25%	25%
HU-III						
Sell	25%	0.1%	1%	10%	0.1%	0.1%
Ann1	25%	ns	ns	ns	25%	ns
LE-IV						
Sell	ns	0.1%	0.1%	25%	0.1%	0.1%
Ann1	ns	ns	ns	ns	10%	25%
SH-VI						
Sell	ns	0.1%	0.1%	1%	0.1%	0.1%
Ann1	25%	25%	25%	25%	25%	ns
WH-V						
Sell	ns	1%	0.1%	0.1%	0.1%	0.1%
Ann1	ns	25%	25%	1%	25%	25%

Ann1 Annexation Thresholds
Sell Homogeneity Thresholds

However, the effects are consistent and the cumulative effects of both the annexation and the homogeneity parameters are statistically significant when the data sets are considered together.

iv Classifier Comparisons

Differences in the results achieved by the Supervised ECHO, the Nonsupervised ECHO, and the perpoint classifiers over the Landsat data sets will be examined in two ways. First, the classifiers will be compared by examining the results of all 768 observations of the Supervised ECHO classifier, versus all 81 observations of the Nonsupervised ECHO classifier, versus 10 observations of the perpoint classifier over the Landsat data. Then, the optimal results for each classifier, on each data set will be compared.

Comparison of all Landsat Observations

Table 2a-7 presents the effects of classifier on each of the six dependent variables. As can be seen from this table, the choice of classifier has a significant statistical effect on each of the six dependent variables. Table 2a-8 summarizes the results of pair wise comparisons of the classifiers. By examining this table, we may identify statistically significant differences between specific pairs of classifiers. By examining the average response over all ECHO parameters settings measured for the three classifiers, the following conclusions may be drawn:

- * The Supervised ECHO classifier requires significantly less computer time than either the perpoint or Nonsupervised ECHO classifiers. There is not, however, a significant difference between the computer time required to classify an area by a perpoint classifier and the computer time required to classify the same area by the Nonsupervised ECHO classifier.
- * The field center pixel performance of the Nonsupervised ECHO classifier is significantly lower than the field center pixel performance of the perpoint classifier and the field center pixel performance of the Supervised ECHO classifier, when all parameters settings are considered. The field center pixel performance achieved by the Supervised ECHO classifier is not significantly different from the field center pixel performance of the perpoint classifier when the results of the 24 Supervised ECHO classifications produced with the differing parameters settings for each data set are averaged.
- * The full field performance of the Nonsupervised classifier is significantly lower than the full field performance of the Supervised ECHO and perpoint classifiers. There is no statistical difference between the performance of the Supervised ECHO classifier and the perpoint classifier when the results from all 24 ECHO parameter settings of the Supervised ECHO processor are averaged and compared to the perpoint result.

Table 2a-7

Overall Landsat Comparisons:
Comparison of Perpoint, Supervised ECHO,
and Nonsupervised ECHO Classifiers

<u>Variable</u>	<u>Degrees of Freedom</u>	<u>F-Value</u>
CPU Time	2,16	21.64***
FCP Performance	2,16	14.41***
Full Field Performance	2,16	12.85***
Training Performance	2,16	6.63**
RMS Error	2,16	4.16*
Classification Variability	2,16	10/54**

Significance Level

*	5%
**	1%
***	.1%

Table 2a-8

Overall Landsat Comparisons:
Comparison of Perpoint, Supervised ECHO, and
Nonsupervised ECHO Classifiers
(859 Observations)

1 = Supervised ECHO (768 Observations)
2 = Nonsupervised ECHO (81 Observations)
3 = Perpoint (10 Observations)

Range Tests
(T-tests with 17 degrees of Freedom)

CPU Time	1	<u>3</u>	<u>2</u>
1,2		6.61***	
1,3		2.11*	
2,3		.30	
Field Center Pixel Performance	2	<u>3</u>	<u>1</u>
1,2		6.59***	
1,3		.04	
2,3		2.26*	
Full Field Performance	2	<u>1</u>	<u>3</u>
1,2		6.12***	
1,3		.06	
2,3		2.19*	
Training Field Performance	<u>2</u>	<u>3</u>	<u>1</u>
1,2		4.25***	
1,3		.01	
2,3		1.47	
RMS Proportion Error	<u>3</u>	<u>1</u>	<u>2</u>
1,2		3.35**	
1,3		.13	
2,3		1.29	
Classification Variability	<u>1</u>	<u>2</u>	<u>3</u>
1,2		1.67	
1,3		4.99***	
2,3		4.17***	

- * The training field performance of the Supervised ECHO classifier is significantly higher than the training field performance of the Nonsupervised ECHO classifier. The training field performance of the perpoint classifier is not significantly different from the training field performance of the Nonsupervised ECHO classifier; nor is the training field performance perpoint classifier significantly different from the training field performance of the Supervised ECHO classifier.
- * The RMS proportion estimate error of the Nonsupervised ECHO classifier is significantly larger than the RMS proportion estimate error of either the perpoint or the Supervised ECHO classifiers. Again, the RMS proportion errors for the perpoint and the Supervised ECHO classification results are not significantly different.
- * Both the Nonsupervised and the Supervised ECHO classifiers have significantly lower variability of classification results than the perpoint classifier. The variability of the classification results for the two ECHO algorithms are not significantly different, however.

These results indicate that; with random selection of homogeneity and annexation parameters, both ECHO classifiers perform better than the perpoint classifier only with respect to classification variability, and the Supervised ECHO classifier will require less time than the perpoint or the Nonsupervised ECHO classifier. On the other hand, with random selection of parameter settings, the Nonsupervised ECHO classifier has inferior performance with respect to field center pixel accuracy, full field accuracy and RMS proportion error to those of the Supervised ECHO and the perpoint classifiers.

Comparison of the Optimal Landsat Results

Table 2a-9 summarizes the optimal results for each dependent variable of each classifier for each Landsat data set. In addition, this table lists the difference between each pair of classifiers to be examined. The difference between the conclusions reached in this section and the conclusions reached in the previous section is that, in the previous section, the results for all the homogeneity and annexation parameter settings were considered for the Nonsupervised and Supervised ECHO classifiers and these classifiers were then compared to each other and the perpoint classifier; in this section only the results yielded by the optimal parameter settings of the Supervised and Nonsupervised ECHO classifiers are considered.

Table 2a-10 summarizes the results of paired T-tests between the optimal performances of each pair of classifiers over the Landsat data sets.

TABLE 2a-9
OPTIMAL RESULTS OF LANDSAT DATA SETS

Data Set	Variable	Perpoint	Super.	Nonsup.	Perpoint -Sup.	Perpoint -Nonsup	Nonsup. -Sup.
GRAHAM	CPU	121.00	74.00	154.00	47.00	-33.00	80.00
	FCP	89.90	93.90	95.10	-4.00	-5.20	1.20
	FF	87.40	90.00	92.00	-2.60	-4.60	2.00
	TRAIN	90.40	98.50	94.50	-8.10	-4.10	-4.00
	RMS	1.20	.10	.10	1.10	1.10	0.00
	VAR	20.85	10.80	10.49	10.05	10.36	-.31
GRANT	CPU	122.00	98.00	149.00	24.00	-27.00	51.00
	FCP	64.50	68.10	60.80	-3.60	3.70	-7.30
	FF	66.40	70.80	64.90	-4.40	1.50	-5.90
	TRAIN	56.20	58.40	54.00	-2.20	2.20	-4.40
	RMS	6.50	.10	2.28	-1.60	4.22	-5.82
	VAR	24.62	8.60	12.81	16.02	11.82	4.20
KEARNY	CPU	389.00	100.51	205.00	288.49	184.00	104.49
	FCP	65.00	65.40	51.00	-0.40	14.00	-14.40
	FF	63.90	64.40	52.30	-0.50	11.60	-12.10
	TRAIN	67.60	68.20	45.20	-0.60	22.40	-23.00
	RMS	12.30	12.28	8.00	.02	4.30	-4.28
	VAR	44.12	16.16	11.36	27.96	32.76	-4.80
HASKELL	CPU	396.00	166.82	223.00	229.18	173.00	56.18
	FCP	67.30	77.20	65.40	-9.90	1.90	-11.80
	FF	67.00	76.40	64.50	-9.40	2.50	-11.90
	TRAIN	77.70	88.50	63.80	-10.80	13.90	-24.70
	RMS	3.50	.98	.10	2.52	3.40	-0.88
	VAR	45.73	14.07	13.55	31.66	32.18	-0.52
LE-IV	CPU	284.00	180.22	306.00	103.78	-22.00	125.78
	FCP	50.70	57.20	55.10	-6.50	-4.40	-2.10
	FF	50.60	55.70	54.20	-5.10	-3.60	-1.50
	TRAIN	69.00	73.00	60.60	-4.00	8.40	-12.40
	RMS	16.20	19.37	23.78	-3.17	-7.58	4.41
	VAR	46.38	8.54	3.29	37.84	43.09	-5.25
HU-III	CPU	852.00	480.00	696.00	372.00	156.00	216.00
	FCP	59.10	61.70	61.90	-2.60	-2.80	.20
	FF	59.50	59.70	52.00	-0.20	7.50	-7.70
	TRAIN	85.20	90.80	40.00	-5.60	45.20	-50.80
	RMS	14.70	11.59	6.83	3.11	7.87	-4.76
	VAR	48.90	18.90	17.80	30.00	31.10	-1.10

TABLE 2a-9 (cont'd)
OPTIMAL RESULTS OF LANDSAT DATA SETS

Data Set	Variable	Perpoint	Super.	Nonsup.	Perpoint -Sup.	Perpoint -Nonsup.	Nonsup. -Sup.
SH-VI	CPU	526.00	296.75	473.00	229.25	53.00	176.25
	FCP	66.90	63.02	50.60	3.80	16.30	-12.42
	FF	62.40	60.20	50.10	2.20	12.30	-10.10
	TRAIN	59.20	65.70	53.10	-6.50	6.10	-12.60
	RMS	21.60	20.50	19.24	1.10	2.36	-1.26
	VAR	48.45	13.09	9.53	35.36	38.92	-3.56
WH-V	CPU	458.00	239.13	425.00	218.87	33.00	185.87
	FCP	74.60	75.50	57.30	-0.90	17.30	-18.20
	FF	70.60	70.51	57.60	.09	13.00	-12.91
	TRAIN	87.80	91.00	65.90	-3.20	21.90	-25.10
	RMS	6.90	7.20	23.13	-0.30	-16.23	15.93
	VAR	47.87	19.30	10.75	28.57	37.12	-8.55
LIVSTON	CPU	513.00	253.00		260.00		
	FCP	71.30	75.20		-3.90		
	FF	68.90	69.40		-0.50		
	TRAIN	85.10	88.20		-3.10		
	RMS	12.80	7.00		5.80		
	VAR	53.07	18.00		35.07		
FA-V	CPU	348.00	198.88	325.50	149.12	22.50	126.62
	FCP	86.60	89.20	89.50	-2.60	-2.90	.30
	FF	78.40	80.10	74.30	-1.70	4.10	-5.80
	TRAIN	72.20	78.00	88.50	-5.80	-16.30	10.50
	RMS	4.70	5.03	16.05	-0.33	-11.53	-11.02
	VAR	44.30	18.28	18.80	26.02	25.50	.52

CPU times are in seconds.

Field center pixel (FCP), full field (FF), and training field (TRAIN) performances, RMS proportion estimate error, and classification variabilities (VAR) are in percentages.

TABLE 2a-9 (cont'd)
OPTIMAL RESULTS OF LANDSAT DATA SETS

Data Set	Variable	Perpoint	Super.	Nonsup.	Perpoint -Sup.	Perpoint -Nonsup.	Nonsup. -Sup.
SH-VI	CPU	526.00	296.75	473.00	229.25	53.00	176.25
	FCP	66.90	63.02	50.60	3.80	16.30	-12.42
	FF	62.40	60.20	50.10	2.20	12.30	-10.10
	TRAIN	59.20	65.70	53.10	-6.50	6.10	-12.60
	RMS	21.60	20.50	19.24	1.10	2.36	-1.26
	VAR	48.45	13.09	9.53	35.36	38.92	-3.56
WH-V	CPU	458.00	239.13	425.00	218.87	33.00	185.87
	FCP	74.60	75.50	57.30	-0.90	17.30	-18.20
	FF	70.60	70.51	57.60	.09	13.00	-12.91
	TRAIN	87.80	91.00	65.90	-3.20	21.90	-25.10
	RMS	6.90	7.20	23.13	-0.30	-16.23	15.93
	VAR	47.87	19.30	10.75	28.57	37.12	-8.55
LIVSTON	CPU	513.00	253.00		260.00		
	FCP	71.30	75.20		-3.90		
	FF	68.90	69.40		-0.50		
	TRAIN	85.10	88.20		-3.10		
	RMS	12.80	7.00		5.80		
	VAR	53.07	18.00		35.07		
FA-V	CPU	348.00	198.88	325.50	149.12	22.50	126.62
	FCP	86.60	89.20	89.50	-2.60	-2.90	.30
	FF	78.40	80.10	74.30	-1.70	4.10	-5.80
	TRAIN	72.20	78.00	88.50	-5.80	-16.30	10.50
	RMS	4.70	5.03	16.05	-0.33	-11.53	-11.02
	VAR	44.30	18.28	18.80	26.02	25.50	.52

CPU times are in seconds.

Field center pixel (FCP), full field (FF), and training field (TRAIN) performances, RMS proportion estimate error, and classification variabilities (VAR) are in percentages.

Table 2a-10

Comparison of the Optimal Landsat Results for
the Supervised ECHO, Nonsupervised ECHO and Perpoint Classifier

Variable	Perpoint versus Supervised	Perpoint versus Nonsupervised	Nonsupervised versus Supervised
<u>Observations</u>	10	9	9
CPU			
T	5.53	2.03	6.41
Significance Level	.1%	5%	.1%
FCP			
T	2.65	1.37	2.94
Significance Level	1%	NS	1%
FF			
T	2.1	2.22	4.31
Significance Level	5%	5%	1%
Training			
T	5.24	1.88	2.81
Significance Level	.1%	5%	5%
RMS Error			
T	1.03	.48	.33
Significance Level	NS	NS	NS
Classification Variability			
T	10.08	7.67	1.71
Significance Level	.1%	.1%	NS

At optimal parameter settings, both the Nonsupervised and Supervised ECHO classifiers require significantly less CPU time than the perpoint classifier. This is true even though the Nonsupervised classifier was run with the "MAP" option specified for the Landsat data. In addition, the Supervised ECHO classifier requires significantly less time than the Nonsupervised ECHO classifier for these data sets. The specification of the "MAP" option for the Nonsupervised classifier contributed to this result.

The Supervised classifier demonstrates a significantly superior field center pixel performance to either the perpoint or the Nonsupervised ECHO classifier. However, the Nonsupervised classifier, which identifies objects without the benefit of class statistics, demonstrates no statistically significant difference from the perpoint classifier with respect to field center pixel accuracy, at optimal annexation and homogeneity parameter settings.

The full field performance of the Supervised ECHO classifier is significantly superior to that of the perpoint and Nonsupervised ECHO classifiers. On the other hand, the full field performance of the Nonsupervised ECHO classifier at its optimal parameter settings is still significantly inferior to that of the perpoint classifier.

For the Landsat data sets, the Supervised ECHO classifier has a statistically significant advantage in training performance to both the Nonsupervised ECHO and the perpoint classifiers. Again, even at optimal parameter settings the Nonsupervised ECHO classifications are at a statistically significant disadvantage when compared to the perpoint results.

There are no statistically significant differences between the perpoint, the Supervised ECHO and the Nonsupervised ECHO classifiers with respect to RMS proportion estimate errors.

While not being significantly different from each other, the results of both the Supervised and the Nonsupervised ECHO classifiers at their optimal parameter settings demonstrate significantly less classification variability than the perpoint results.

b. Simulated Thematic Mapper Results

The analyses of variance for the effects of the Nonsupervised ECHO parameters on the six simulated Thematic Mapper data sets (two resolutions from Williams County, ND, four resolutions from Finney County, Kansas) are presented in Table 2a-11. Results for the six data sets considered together are presented in Table 2a-12.

1 Cell Width Parameter

For the Landsat and aircraft data sets the Nonsupervised ECHO cell width parameter was set to two in all cases. For those data sets the cell width setting of two was selected to economize on the CPU time required to perform the verification tests based on the observation that, for Supervised ECHO results, larger cell width parameters did not improve ECHO performances. The six simulated Thematic Mapper data sets were classified by the Nonsupervised ECHO classifier at cell widths of two and three. There were several statistically significant effects of the cell width parameter on the Nonsupervised ECHO classifications of simulated Thematic Mapper data:

- * When the cell width parameter is set to three, less CPU time is required than when it is set to two (significant at a 10% confidence level).
- * The training performance at cell width three is significantly poorer than at cell width two (5% confidence level).
- * The RMS proportion estimate error is less at cell width three than at cell width two (10% confidence level).
- * The classification variability is less for cell width two than cell width three (10% confidence level).

The fact that the proportion estimates are better for cell width three than for cell width two for the simulated Thematic Mapper data indicates that the partitioning using the larger cell size may, indeed, be of value with the Nonsupervised ECHO processor.

Because the likelihood that a cell will be homogeneous when it contains nine scene elements is smaller than the likelihood that a cell containing only four scene elements will be homogeneous, it makes sense that, for a given homogeneity setting, more cells at cell width three, than at cell width two, would be identified as singular and split with constituent points classified individually. This situation may explain why the classification variability at cell width two is smaller than the classification variability at cell width three. On the other hand, when large homogeneous areas are present in a scene, fewer calculations of the classification equation will be necessary at larger cell widths. Thirty-six pixels in 2 by 2 pixel cells require nine calculations of class probability for each spectral class while, in 3 by 3 pixel cells, only four calculations of class probability for each spectral class are required.

Table 2a-11

Results of Individual Thematic Mapper Data Set
Parameter Anovas
(108 Observations)

	CPU	FCP	FULL	TRAIN	RMS	VARI	DF
1730							
Celw	1%	ns	5%	25%	1%	1%	1,4
Sell	0.1%	5%	0.1%	1%	1%	0.1%	2,4
Annl	1%	25%	25%	ns	10%	0.1%	2,4
Celw X Sell	1%	ns	10%	10%	ns	0.1%	2,4
Celw X Annl	1%	ns	ns	ns	25%	5%	2,4
Sell X Annl	25%	ns	ns	ns	10%	5%	4,4
1740							
Celw	0.1%	5%	10%	ns	1%	1%	1,4
Sell	0.1%	5%	1%	1%	0.1%	0.1%	2,4
Annl	1%	25%	25%	5%	1%	0.1%	4,4
Celw X Sell	5%	ns	ns	ns	5%	0.1%	2,4
Celw X Annl	5%	ns	ns	25%	ns	5%	2,4
Sell X Annl	5%	25%	25%	25%	1%	1%	4,4
3730							
Celw	0.1%	ns	ns	ns	ns	ns	1,4
Sell	0.1%	0.1%	0.1%	0.1%	0.1%	0.1%	2,4
Annl	0.1%	5%	5%	25%	ns	5%	4,4
Celw X Sell	1%	ns	ns	ns	10%	5%	2,4
Celw X Annl	1%	25%	10%	25%	25%	25%	2,4
Sell X Annl	5%	5%	5%	25%	ns	25%	4,4
3740							
Celw	1%	ns	10%	ns	25%	5%	1,4
Sell	0.1%	ns	0.1%	0.1%	0.1%	0.1%	2,4
Annl	0.1%	ns	5%	ns	25%	1%	2,4
Celw X Sell	1%	ns	1%	ns	0.1%	1%	2,4
Celw X Annl	1%	ns	ns	ns	ns	5%	2,4
Sell X Annl	5%	ns	25%	ns	ns	10%	4,4
3750							
Celw	5%	0.1%	0.1%	0.1%	0.1%	0.1%	1,4
Sell	0.1%	0.1%	0.1%	0.1%	0.1%	0.1%	2,4
Annl	1%	ns	25%	5%	10%	10%	2,4
Celw X Sell	5%	1%	5%	5%	5%	1%	2,4
Celw X Annl	1%	ns	ns	ns	ns	5%	2,4
Sell X Annl	10%	25%	25%	10%	25%	25%	4,4

Table 2a-11 (Continued)

	CPU	FCP	FULL	TRAIN	RMS	VARI	DF
3760							
Celw	25%	ns	ns	ns	10%	ns	1,4
Sell	0.1%	1%	1%	0.1%	1%	5%	2,4
Annl	1%	ns	ns	ns	ns	ns	2,4
Celw X Sell	5%	ns	ns	ns	ns	ns	2,4
Celw X Annl	5%	25%	25%	25%	ns	ns	2,4
Sell X Annl	5%	ns	ns	ns	ns	ns	4,4

Table 2a-12

OVERALL THEMATIC MAPPER RESULTS
Significant Effects of Parameters on Dependent Variables
(108 Observations)

Parameters	Degrees of Freedom	Variables					
		CPU Time	FCP Perform.	Full Field Perform.	Training Perform.	RMS Error	Classi. Vari.
Resolution	3,1	8.35	57.90+	19.44	509.24*	5.41	2.85
Cell Width	1,1	91.73+	3.34	0.76	345.90*	119.46+	72.44+
Homogeneity Threshold	2,2	25.51*	32.50*	16.26+	164.70**	68.63*	73.41*
Annexation Threshold	2,2	178.21**	4.05	3.45	13.46+	9.78+	65.95*
Resolution X Cell Width	3,1	92.01+	14.51	4440.00*	64.71+	19.90	38.07
Resolution X Homogeneity	6,2	3.68	50.21*	22.59*	12.70+	43.59*	48.58*
Resolution X Annexation	6,2	.68	1.72	3.05	.31	8.07	1.87
Cell Width X Homogeneity	2,2	45.59*	479.36**	7.68	44.74*	5.83	12.87+
Cell Width X Annexation	2,2	30.27*	19.66*	5.74	11.05+	286.00**	63.04**
Homogeneity X Annexation	4,4	227.36***	2.89	3.2	29.19**	4.21+	96.25***

Significance Level

+	10%
*	5%
**	1%
***	.1%

Table entries are for F-values

There are several interaction effects of cell width and resolution. As resolution increases, the effect of the cell width parameter on CPU time decreases (10% confidence level) and the effects of the cell width parameter on the full field and the training performances decrease (10% and 5% confidence levels, respectively). As resolution elements become larger, fields will be sampled by fewer pixels. Therefore, the number of truly homogeneous cells at a given cell size will decrease. Since there will be fewer homogeneous cells at 60 meter resolution than at 30 meter resolution, more cell splitting will take place and, hence, the effects of aggregating pixels into cells will be smaller at 60 meter resolution than at 30 meter resolution. This effect expands as the cell size becomes larger since the number of homogeneous 3 by 3 cells will shrink faster than the number of homogeneous 2 by 2 cells as the resolution elements become larger.

The interaction of the cell width parameter and the homogeneity parameter is statistically significant with respect to CPU time, field center pixel performance, training field performance, and variability of classification results. In all cases, as the cell width parameter increases, the effect of the homogeneity parameter increases. This results is logical since, as the cell become larger, the number of pixels affected by the outcome of each homogeneity test is larger.

There are also significant cell width-annexation threshold interactions. As the cell width becomes larger, the effect of the annexation parameter becomes smaller for CPU time, training field performance, field center pixel performance, RMS proportion estimate error, and classification variability. This results reflects the facts that as the partition becomes larger, there will be fewer and fewer homogeneous cells, and that, as the partition becomes larger, it becomes less likely that a neighboring homogeneous cell will be contained in a single object.

11 Resolution

As resolution increases, both field center pixel and training field performances show a statistically significant decrease. The 50 meter Thematic Mapper data is probably responsible for this circumstance. Both the field center pixel and the training field performances on the 50 meter data sets are significantly below the performance levels of the 40 and the 60 meter data sets. Reasons for this degradation in performance at 50 meters are unclear. One would expect a tailing off of classification accuracy as the size of the cell approaches the size of the objects on the ground. However, in this data set, even at 60 meter resolution, cell sizes do not approach the size of the agricultural fields.

The interaction of the resolution with the homogeneity parameter is statistically significant for the field center pixel performance, the full field performance, the training performance, the RMS proportion estimate error, and the classification variability. For the 30 meter resolution the field center pixel and the full field performances improve as the cell homogeneity parameter increases from .05 to .10, and then fall somewhat as the homogeneity parameter goes from .10 to .25. At the larger resolutions, however, these variables fall steadily as the homogeneity parameter is increased. The effects of the homogeneity parameter increase for: the RMS proportion estimate error, the training field performance, and the classification variability as the resolution element size increases.

Interaction of the resolution and the annexation thresholds is not significant for any of the six dependent variables monitored.

iii Homogeneity Threshold

The homogeneity threshold has a significant effect on all six dependent variables. As the homogeneity threshold increases:

- * the CPU time required to perform the Nonsupervised ECHO classification decreases,
- * the field center pixel performance decreases,
- * the full field performance decreases,
- * the training field performance decreases,
- * the RMS proportion estimate error increases, and
- * the classification variability decreases.

Figures 2a-21 through 2a-26 graph the effects of the homogeneity and annexation parameters on the six dependent variables for 2 by 2 pixel cells. Unlike the Landsat results (See figure 2a-15), for the simulated Thematic Mapper data sets, the effect of homogeneity parameter on CPU time is statistically significant. The "MAP" option was not used on the simulated Thematic Mapper data. Therefore, there was no data replacement when homogeneous objects were identified in the field extraction phase, and consequently, the CPU time required goes down as the number of homogeneous cells increases due to the reduction in the number of passes through the classification equation made possible by classifying a field, rather than a point at a time.

The results of the field center pixel performance, full field performance, training field performance, and RMS proportion estimate error measurements indicate that for simulated Thematic Mapper data, at resolutions above 30 meters, homogeneity parameter specifications of .05 or less are appropriate.

Interaction between the homogeneity parameter and the annexation thresholds is significant for CPU time, training field performance, RMS proportion estimate error, and classification variability. As the homogeneity parameter increases, the effects of the annexation also increase for each of the dependent variables listed above. Since annexation can take place only when adjoining cells are homogeneous and since the number of homogeneous fields increases as the homogeneity parameter increases, it makes sense for the annexation parameter to have a greater effect when the homogeneity parameter is larger.

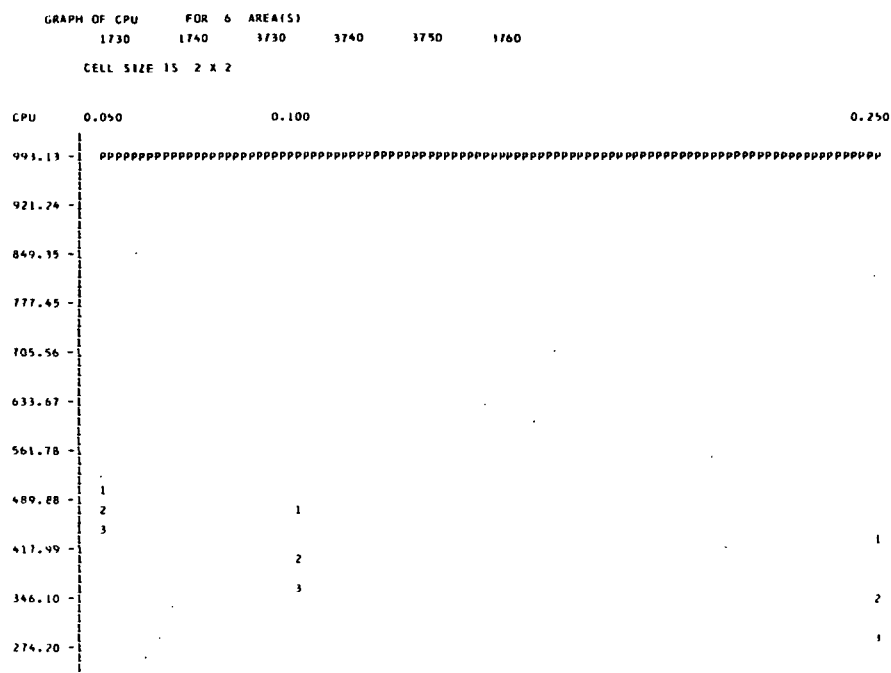


Figure 2a-21

Nonsupervised ECHO CPU Requirements
for the Simulated Thematic Mapper Data Sets

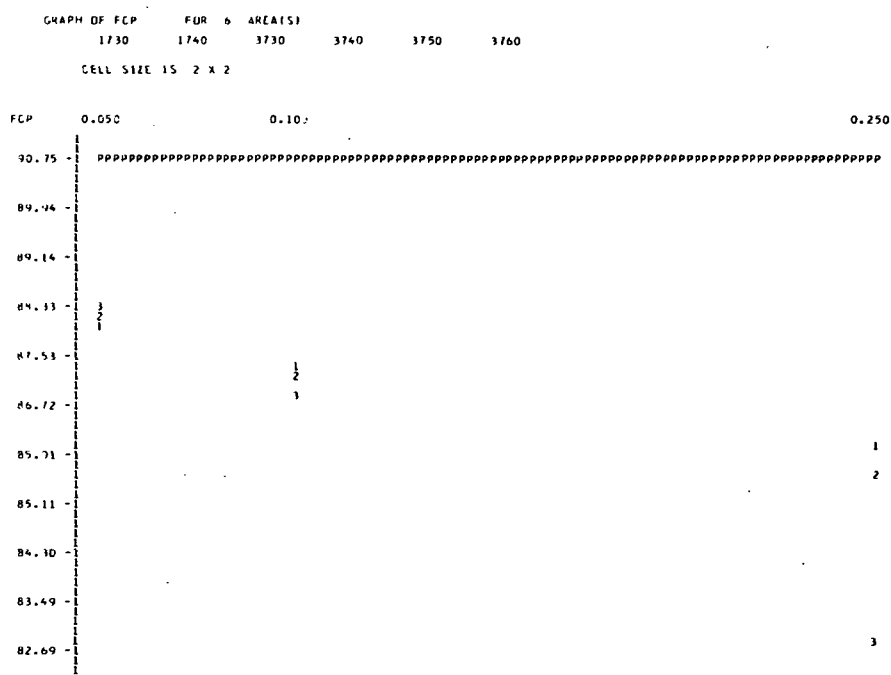


Figure 2a-22

Nonsupervised ECHO Field Center Pixel Performance
for the Simulated Thematic Mapper Data Sets

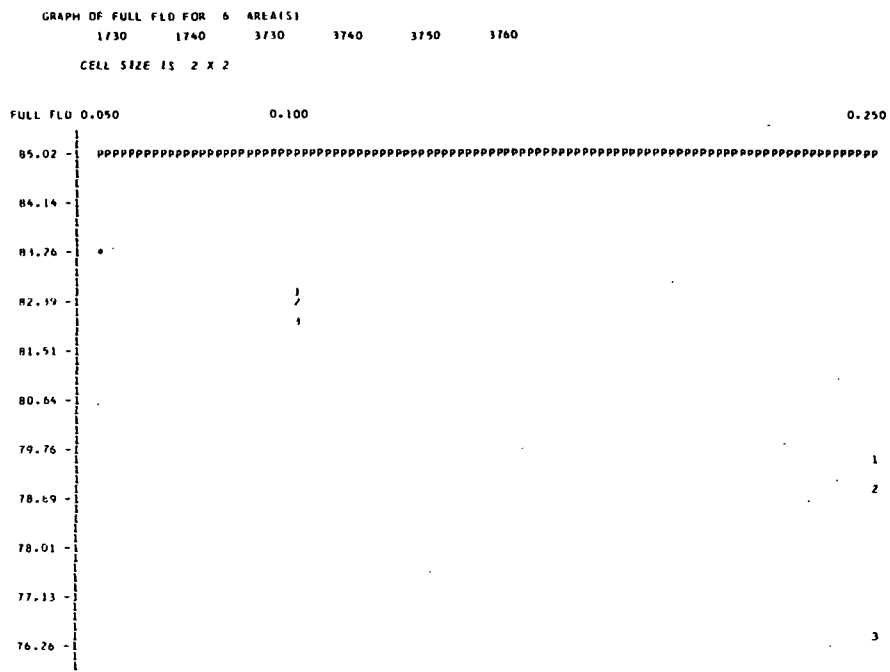


Figure 2a-23

Nonsupervised ECHO Full Field Performance
for the Simulated Thematic Mapper Data Sets

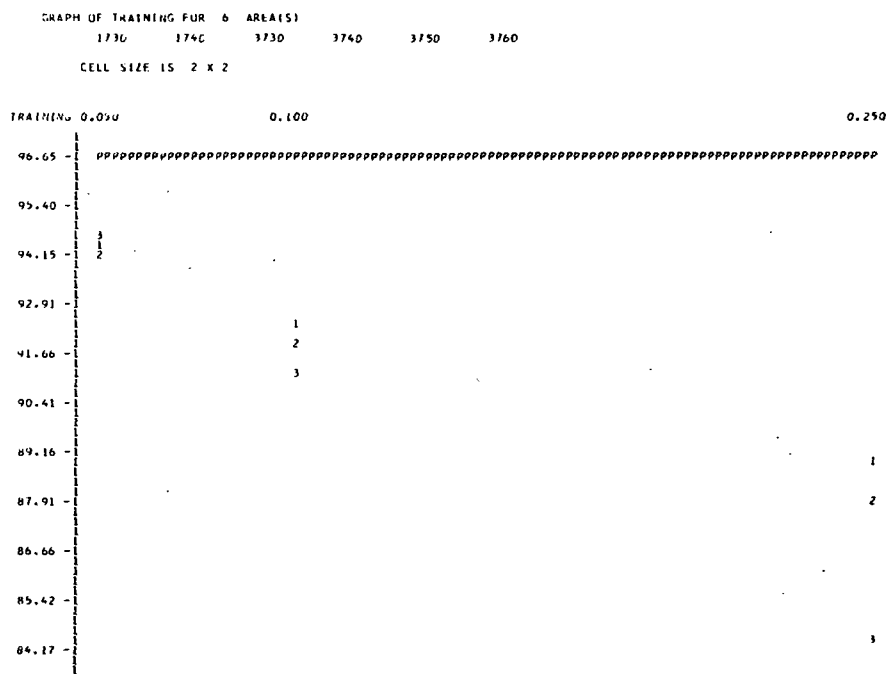


Figure 2a-24

Nonsupervised ECHO Training Field Performance
for the Simulated Thematic Mapper Data Sets

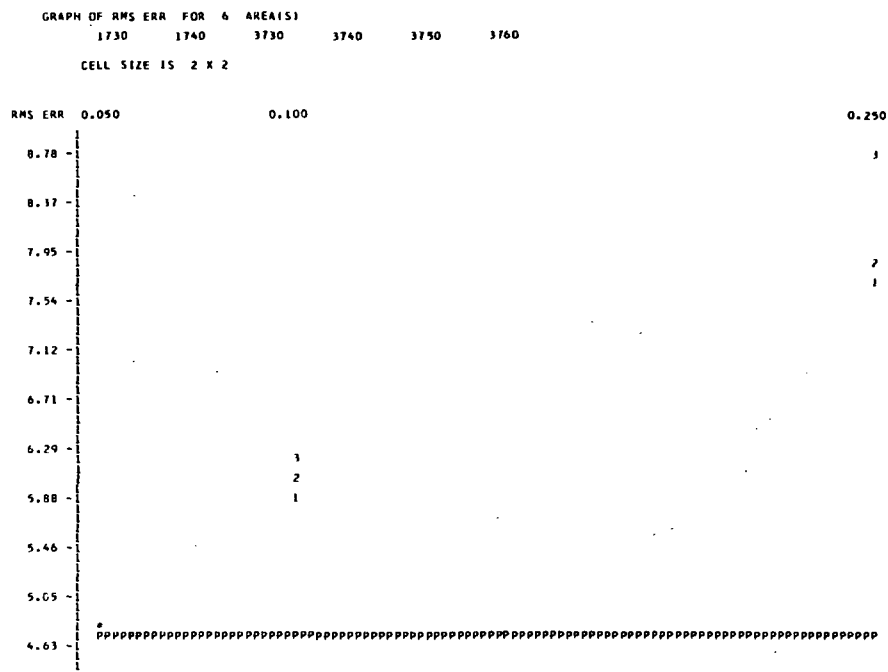


Figure 2a-25

Nonsupervised ECHO RMS Proportion Estimate Error for the Simulated Thematic Mapper Data Sets

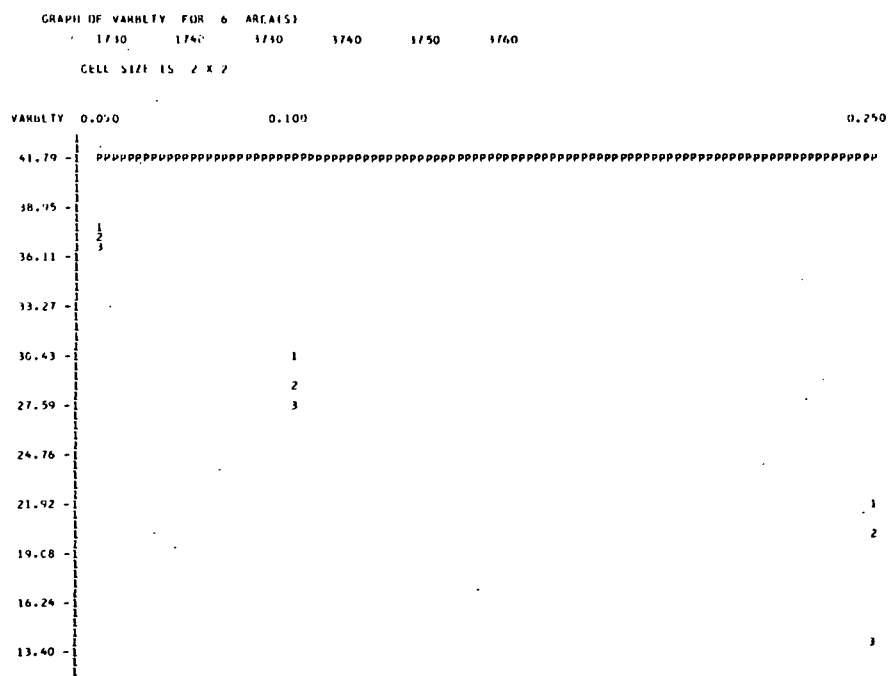


Figure 2a-26

Nonsupervised ECHO Classification Variability for the Simulated Thematic Mapper Data Sets

iv Annexation Threshold

The annexation threshold has a significant effect on the CPU time required to perform classification, on the training performance, on the RMS proportion estimate error, and on the classification variability. As the annexation threshold goes from 10^{-1} to 10^{-3} , the time required to perform a classification decreases, the training performance decreases, the RMS proportion estimate error increases, and the classification variability decreases. This result indicates that for the simulated Thematic Mapper data, annexation parameters in the neighborhood of 10^{-1} are most appropriate.

v Classifier Comparisons

Differences in the results achieved by the Supervised ECHO, the Non-supervised ECHO, and the perpoint classifiers over the simulated Thematic Mapper data sets will be examined in two ways. First, the classifiers will be compared by examining the results of 8 observations of the perpoint classifier versus the 108 observations of the Nonsupervised ECHO processor versus the 576 observations of the Supervised ECHO processor over the simulated Thematic Mapper data sets. Then, the optimal results for each classifier, on each data set will be compared.

Comparison of all Simulated Thematic Mapper Observations

Table 2a-13 presents the effects of classifier and resolution on the simulated Thematic Mapper data sets for the six dependent variables when all observations are considered for the ECHO classifiers. The effect of the classifier is significant for all six of these variables. Resolution has a statistically significant effect on field center pixel performance, full field performance, and classification variability. The interaction of the classifier and the resolution is statistically significant for training field performance, field center pixel performance, full field performance, RMS proportion estimate error, and classification variability. Table 2a-14 illustrates some of the effects of the three classifiers. The CPU time required by the Nonsupervised classifier is significantly less than the CPU time required by either the Supervised ECHO or the perpoint classifiers. On the other hand, the Nonsupervised ECHO classifier demonstrates inferior field center pixel, full field, and training field performances when all ECHO observations are considered.

The poor performance of the Nonsupervised ECHO processor on the 50 meter data may be responsible for many of the significant effects. By examining the graphs of the average field center pixel and full field performance in Table 2a-14, it can be seen that the 50 meter resolution is causing the significant interaction between resolution and classifier and the statistically significant differences between classifiers, for these two variables. The 50 meter results are also causing the significant effect of resolution on field center pixel performance, and the interaction effect between resolution and classifier for the training field performance and classification variability variables.

Table 2a-13

Overall Thematic Mapper Results
692 Observations

<u>EFFECT</u>	<u>DEGREES OF FREEDOM</u>	<u>Performance</u>					
		<u>CPU TIME</u>	<u>FCP</u>	<u>FULL FIELDS</u>	<u>TRAINING FIELDS</u>	<u>RMS PROPORTION</u>	<u>CLASSIFICATION VARIABILITY</u>
CLASSIFIER	2,2	18.72+	229.11**	122.61**	514.46**	80.47**	69.45*
RESOLUTION	3,3	5.02	9.51*	10.35*	2.06	2.90	7.409+
CLASSIFIER X RESOLUTION	6,4	.65	9.94*	6.82*	10.77*	5.37+	7.989*

Confidence Level:

Table entries are F-values

+ 10%
* 5%
** 1%
*** .1%

Table 2a-14

Overall Thematic Mapper
Effects of Classifier on Variables

1 = Supervised ECHO Results
2 = Nonsupervised ECHO Results
3 = Perpoint Results

CPU Time	2	<u>1</u>	<u>3</u>
1,2		3.52*	
1,3		2.48	
2,3		3.42*	

RMS Proportion Error	<u>1</u>	<u>3</u>	<u>2</u>
1,2		4.49**	
1,3		.03	
2,3		1.26	

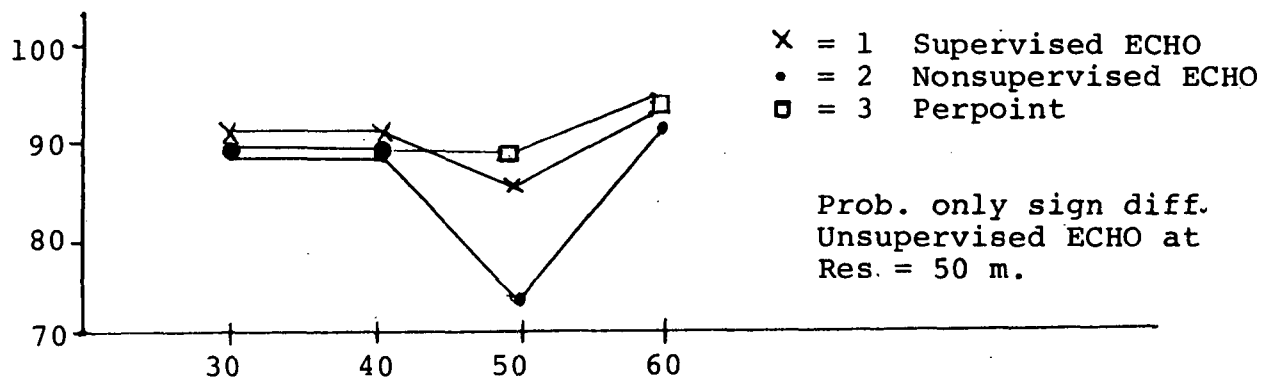
Significance Level

Table entries are T-values

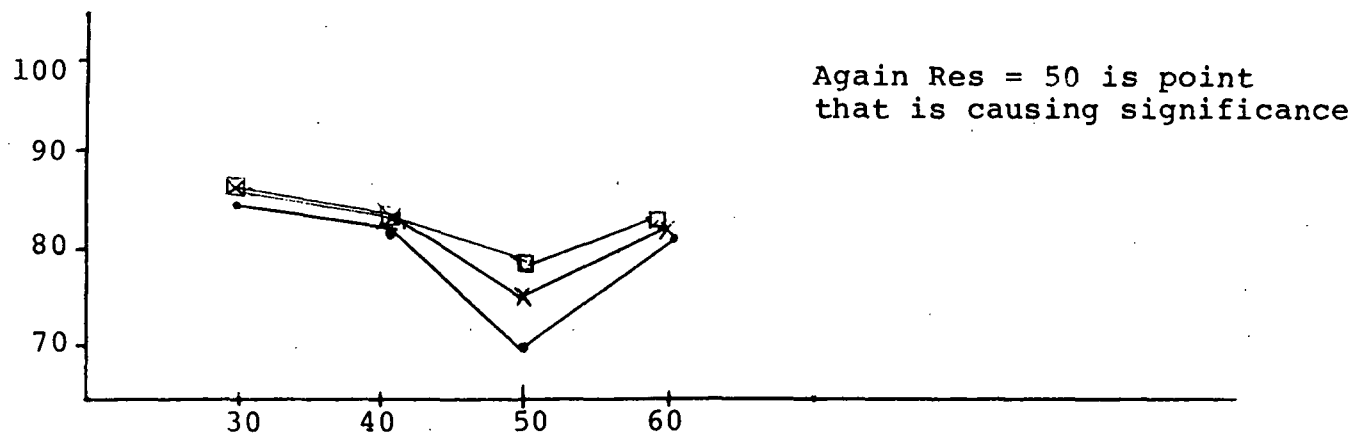
* 5%
** 1%

Table 2a-14 (Continued)

Field Center Pixel Performance



Full Field Performance



Training Field Performance

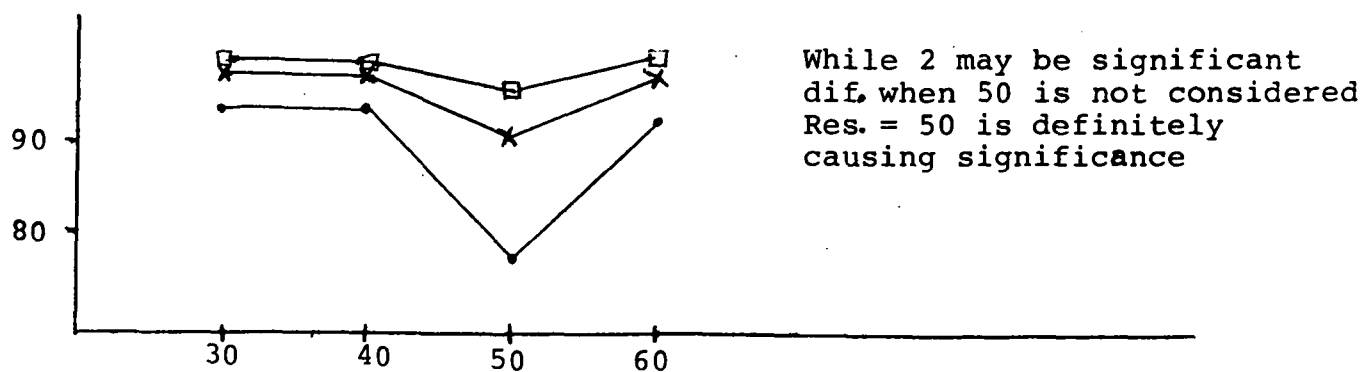
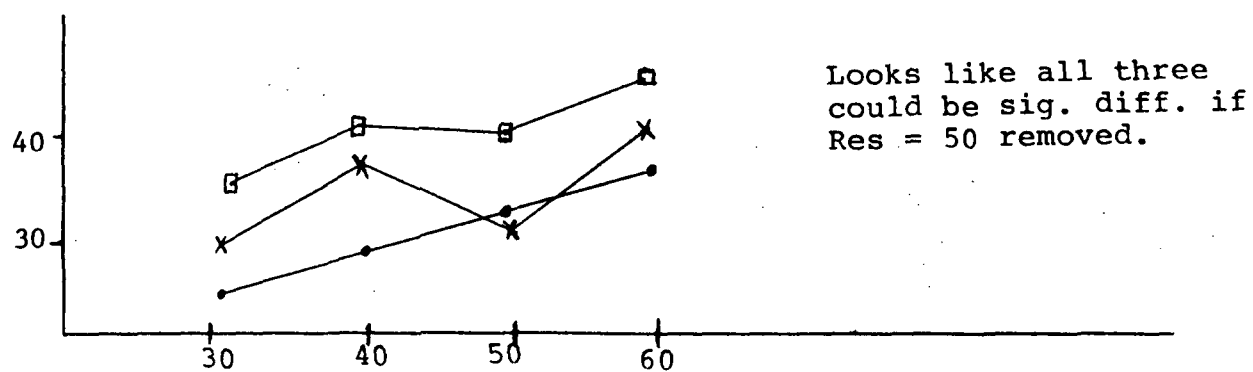


Table 2a-14 (Continued)

Classification Variability



As the resolution elements increase in size, the classification variability increases. This reflects the fact that it becomes less likely that adjacent pixels will be of the same class as the size of resolution elements increase.

There is a statistically significant difference between classifiers with respect to classification variability. The perpoint classification results are more variable than Supervised ECHO classification results which are, in turn, more variable than the Nonsupervised ECHO classification results. The difference between the two ECHO processors indicates that the homogeneity parameters used for the Nonsupervised ECHO tests result in more cells being identified as homogeneous than the homogeneity parameters in the Supervised ECHO experiments. It is possible that for the simulated Thematic Mapper data sets, the optimal homogeneity parameter may be less than the smallest Nonsupervised ECHO homogeneity parameter tested (which was 0.05).

The RMS proportion estimate error for the Supervised ECHO processor is significantly lower than that of the Nonsupervised processor. The proportion estimate error of the perpoint classifier is significantly different from neither the Supervised nor the Nonsupervised ECHO processor.

Comparison of Classifier Optimums

Table 2a-15 presents the optimal dependent variable measurement (lowest CPU, RMS proportion estimate error, classification variability, and highest field center pixel, full field and training field performance) for each of the three classifiers, together with the differences between pairs of optimal responses, for each of the simulated Thematic Mapper data sets. Table 2a-16 presents the results of a paired T-test to identify the significant differences between each pair of classifiers with respect to each dependent variable.

Comparing optimal results, the ECHO classifiers require significantly less CPU time to classify an area than the perpoint classifier requires. The CPU time required to produce a classification by the Nonsupervised and the Supervised ECHO processors are not statistically different, however.

At optimal parameter settings, the Supervised classifier has significantly higher field center pixel and full field classification performances than either the Nonsupervised ECHO or the perpoint classifier. The Nonsupervised ECHO classifier is not significantly different from the perpoint classifier at optimal parameter settings of the Nonsupervised processor with respect to these two variables.

The Supervised ECHO results have training field performance statistically superior to those of the perpoint classifier, which in turn, have the training field performance statistically higher than the results of the Nonsupervised ECHO processor.

The RMS proportion estimate error of the perpoint classifier was not statistically different from either of the ECHO classifiers. The Nonsupervised ECHO classifier had significantly higher proportion estimate error than the Supervised processor, however.

Table 2a-15
OPTIMAL RESULTS OF THEMATIC MAPPER DATA SETS

Data Set	Variable	Perpoint	Super.	Nonsup.	Perpoint -Sup.	Perpoint -Nonsup.	Nonsup. -Sup.
1730	CPU	1170.00	335.20	370.90	834.80	799.10	35.70
	FCP	92.10	94.70	91.20	-2.60	.90	-3.50
	FF	88.10	89.60	87.40	-1.50	.70	-2.20
	TRAIN	97.60	98.50	96.20	-0.90	1.40	-2.30
	RMS	3.40	3.61	3.77	-0.21	-0.37	.16
	VAR	35.93	14.27	12.66	21.66	23.27	-1.61
1740	CPU	732.00	228.94	256.80	503.06	475.20	27.86
	FCP	88.60	91.20	91.70	-2.60	-3.10	.50
	FF	82.60	84.20	85.00	-1.60	-2.40	.80
	TRAIN	97.20	97.70	97.30	-0.50	-0.10	-0.40
	RMS	8.10	7.80	7.98	.30	.12	.18
	VAR	37.68	18.31	13.28	19.37	24.40	-5.03
1750	CPU	436.40	144.97		291.43		
	FCP	88.20	88.60		-0.40		
	FF	74.70	74.90		-0.20		
	TRAIN	98.70	98.90		-0.20		
	RMS	7.60	7.66		-0.06		
	VAR	35.22	15.71		19.51		
1760	CPU	358.80	111.66		247.14		
	FCP	94.50	95.50		-1.00		
	FF	81.00	81.90		-0.90		
	TRAIN	98.70	98.70		0.00		
	RMS	8.50	8.60		-0.10		
	VAR	42.31	22.30		20.01		
3730	CPU	1200.00	353.16	402.50	846.84	797.50	49.34
	FCP	87.50	91.40	89.30	-3.90	-1.80	-2.10
	FF	84.10	87.70	85.80	-3.60	-1.70	-1.90
	TRAIN	97.90	98.60	97.60	-7.00	.30	-1.00
	RMS	7.30	2.40	2.34	4.90	4.96	-0.06
	VAR	34.75	18.47	17.51	16.28	17.24	-0.96
3740	CPU	1222.00	418.43	312.00	803.57	910.00	-106.43
	FCP	91.60	92.80	90.50	-1.20	1.10	-2.30
	FF	86.20	87.60	85.70	-1.40	.50	-1.90
	TRAIN	96.60	98.00	95.90	-1.40	.70	-2.10
	RMS	2.70	2.71	3.17	-0.01	-0.47	.46
	VAR	45.93	27.02	24.63	18.91	21.30	-2.39

Table 2a-15 (cont'd)

OPTIMAL RESULTS OF THEMATIC MAPPER DATA SETS

Data Set	Variable	Perpoint	Super.	Nonsup.	Perpoint -Sup.	Perpoint -Nonsup.	Nonsup. -Sup.
3750	CPU	855.20	233.47	218.60	621.73	636.60	-14.87
	FCP	91.50	91.46	78.10	.04	13.40	-13.36
	FF	85.00	84.68	75.80	.32	9.20	-8.88
	TRAIN	92.90	92.90	84.90	0.00	8.00	-8.00
	RMS	3.80	3.79	7.69	.01	-3.89	3.90
	VAR	46.08	21.20	14.70	24.88	31.38	-6.50
3760	CPU	779.60	322.56	170.40	457.04	609.20	-152.16
	FCP	93.20	93.80	92.10	-0.60	1.10	-1.70
	FF	84.10	84.70	83.30	-0.60	.80	-1.40
	TRAIN	97.70	98.20	96.20	-0.50	1.50	-2.00
	RMS	3.00	2.61	3.53	.39	-0.53	.92
	VAR	50.35	33.97	2.73	16.38	47.62	-31.24

Table 2a-16

Comparison of the Optimal Simulated Thematic Mapper Results
for the Supervised ECHO, Nonsupervised ECHO, and Perpoint Classifiers

Variable	Perpoint versus Supervised	Perpoint versus Nonsupervised	Nonsupervised versus Supervised
Observations			
CPU			
T	6.79	10.87	1.87
Significance Level	.1%	.1%	NS
Field Center Pixel Performance			
T	3.18	.80	1.87
Significance Level	1%	NS	10%
Full Field Performance			
T	2.85	.70	1.93
Significance Level	1%	NS	10%
Training Field Performance			
T	1.58	1.60	2.36
Significance Level	10%	10%	5%
RMS Proportion Estimate Error			
T	1.08	.03	1.52
Significance Level	NS	NS	10%
Classification Variability			
T	19.97	6.21	1.68
Significance Level	.1%	1%	10%

The results of the Nonsupervised ECHO processor were significantly less variable than those of either the perpoint or the Supervised ECHO processor. In addition, the Supervised ECHO classification results were less variable than the perpoint classifier's results at optimal parameter settings.

For the optimal parameter settings of the ECHO classifiers, the performance of the Nonsupervised ECHO processor is superior to the performance of the perpoint classifier only with respect to the CPU time required to perform the classification and the variability of the classification results. The Nonsupervised ECHO processor had training field performance which was worse than the perpoint classifier, for all the parameter settings which were tested (performances would have been the same if no homogeneous cells had been identified). These results indicate that there may be little or not advantage in classifying with the Nonsupervised ECHO processor with respect to classification performances. However, the classification variability results indicate that for the five channel simulated Thematic Mapper data, the homogeneity parameters selected for testing of the Nonsupervised ECHO caused more cells to be identified as homogeneous than were identified by the Supervised processor as homogeneous. All homogeneity parameters for the Nonsupervised ECHO tests over this data set may have been too high. Simply lowering the homogeneity parameter may improve the performance of the Nonsupervised processor. In any case, it seems fair to conclude that the Supervised ECHO processor performs better than both the Nonsupervised ECHO and the perpoint classifiers. Its superiority may be attributable to the use of class statistics in object identification.

c. Aircraft Results

Figures 2a-27 through 2a-31 present the CPU time, the field center pixel performance, the full field performance, the training field performance, and the classification variability achieved by the Nonsupervised ECHO processor for the single aircraft run tested. RMS proportion estimate error values were not calculated.

There are insufficient samples to perform an analysis of variance comparing these results to the single perpoint classification of the area. However, results of T-tests of the probability of each dependent variable's perpoint result falling in the distribution of the nine measurements of that dependent variable for the Nonsupervised ECHO results is presented in Table 2a-17. It appears from these results that, for the aircraft data set, the Nonsupervised ECHO classifier:

- * requires less CPU time than the perpoint classifier,
- * has inferior field center pixel performance to the perpoint classifier, and
- * has superior full field performance when compared to the perpoint classifier.

These results are weak since so few samples are available. On the basis of Landsat and simulated Thematic Mapper results, it seems unlikely that the field center pixel performance is significantly superior for the same data set in which full field performance is significantly inferior.

Table 2a-18 presents the analysis of variance results for the effects of the Nonsupervised ECHO homogeneity and annexation parameters on each of the dependent variables. There are insufficient data points for any effects to be statistically significant at a 10% confidence level.

TABLE 2a-17

T-tests Comparing the Perpoint Results to the Distribution of
Nonsupervised ECHO Results for Six Variables

	CPU Time	FCP	Full Field	Training Field	Classification Variability
Mean of the Nonsupervised Results	279.22	70.53	70.52	85.22	14.94
Standard Devia- tion of Nonsuper- vised Results	39.43	.77	.72	1.01	5.65
Perpoint Results	434.00	71.80	69.50	86.40	22.29
Perpoint result is different than the ECHO result at a 10% confidence level	yes	yes	yes	no	no

TABLE 2a-18

Effects of Nonsupervised ECHO Homogeneity and
Annexation Thresholds on Six Dependent Variables for
One Aircraft Data Set

	CPU	FCP	Full Field	Training Field	Classification Variability
Homogeneity	NS	NS	NS	NS	NS
Annexation	NS	NS	NS	NS	NS

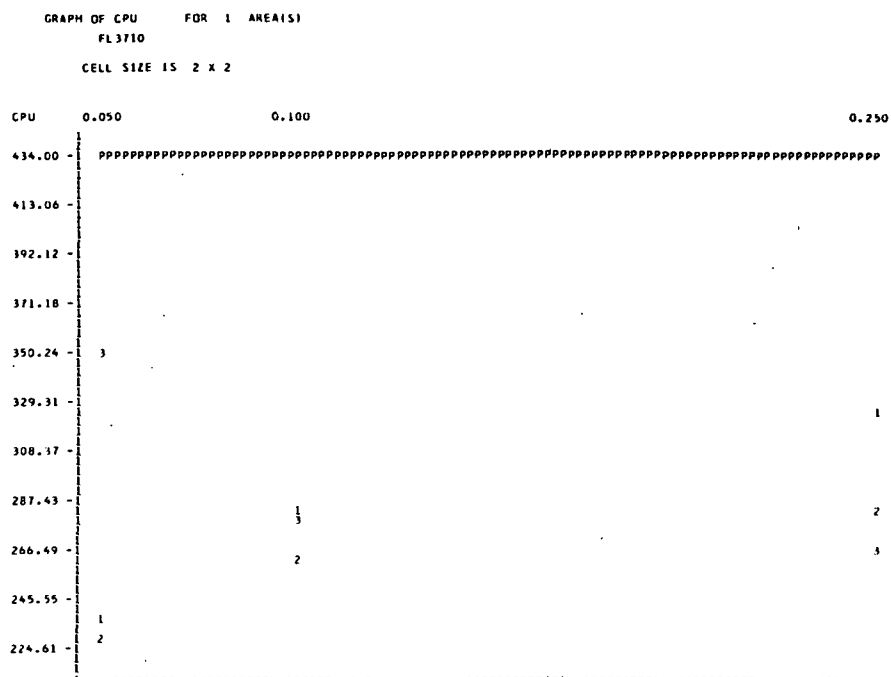


Figure 2a-27

Nonsupervised ECHO CPU Requirements for Aircraft Data Set

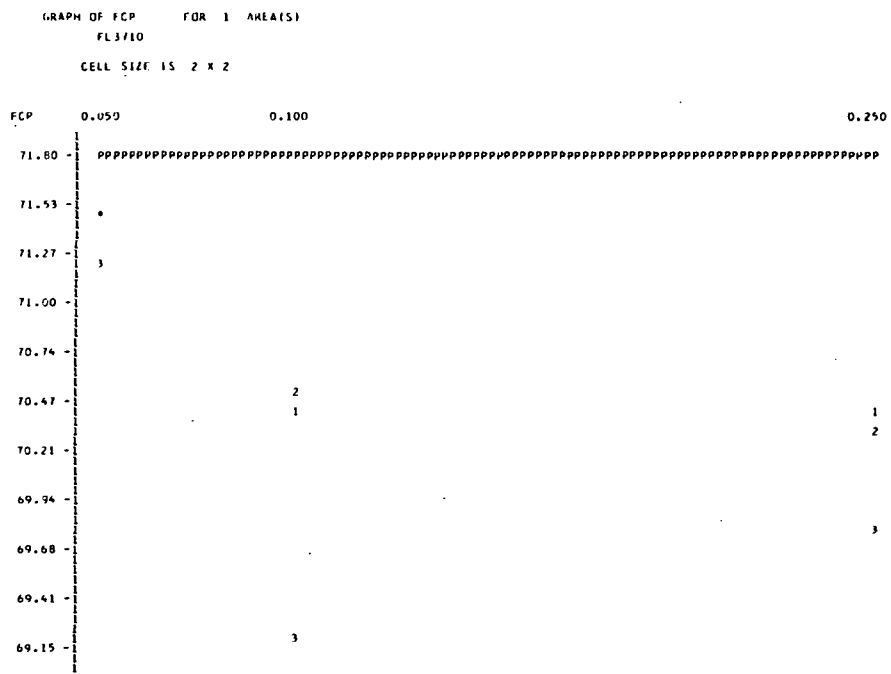


Figure 2a-28

Nonsupervised ECHO Field Center Pixel Performance for Aircraft Data Set

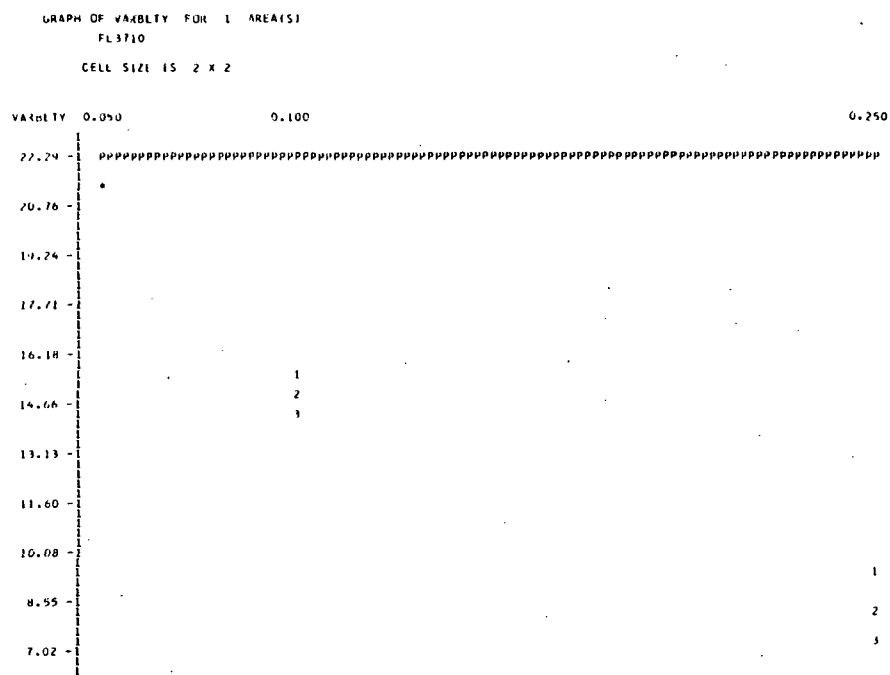


Figure 2a-31

Nonsupervised ECHO Classification Variability
for aircraft data set

Nonsupervised ECHO Object Map Integrity Assessment

The Object Map Integrity Assessment task called for the comparison of the objects found by the Nonsupervised ECHO with the known fields in the scene. Several problems arose which made it impossible to obtain quantitative results. The program written to construct object boundaries from the intermediate results tape took longer than anticipated to debug, and the delay strained the time and personnel resources allotted to this task. The display of the objects in a form suitable for comparison with the field maps of the LACIE segments also encountered difficulties which have not been completely resolved. As a result, no assessment of the object map integrity has been possible to date.

```

10004  END OF INPUT DECK - RUN COMPLETED      (LARSMM)
10050  TOTAL CPU TIME FOR THIS RUN WAS 186.942 SECONDS. (LARSMM)

```

Nonsupervised ECHO Object Map

CONCLUSIONS

Summary

The Supervised ECHO processor (which utilizes class statistics for object identification) successfully exploits the redundancy of states characteristics of sampled imagery of ground scenes to achieve better classification accuracy, reduce the number of classifications required and reduce the variability of classification results. The Supervised ECHO processor requires cell size, cell-to-field annexation, and cell homogeneity parameters, input data, and a class-conditional marginal density statistics deck for both object identification and classification. The improvement in classification performance the Supervised ECHO classifier provides (over a perpoint classifier) expands as the size of the objects expands in terms of numbers of pixels.

The Nonsupervised ECHO processor (which identifies objects without the benefit of class statistics) successfully reduces the number of classifications required and the variability of the classification results. It is unsuccessful in improving classification performance, however. The Nonsupervised ECHO processor runs in two phases, a field extraction phase, then a classification phase. Cell size, cell-to-field annexation, and cell homogeneity parameters, along with input data are required by the field extraction phase of the Nonsupervised processor. The classification phase requires the intermediate results produced by the field extraction phase of the Nonsupervised ECHO processor and a class-conditioned marginal density statistics deck to classify the objects identified in the field extraction phase.

Both ECHO processors provide information which may be of value in the training process. The Supervised processor produces a singular cell map which may be used to assess the adequacy of training for the area classified. Cells are identified as singular by the Supervised processor when the likelihood of the cell belonging to the most likely of the available classes falls below a threshold. Therefore, when a cell is categorized as "singular", it either contains pixels from more than one class or the spectral class of the pixels of the cell is not represented in the available class statistics. Groups of contiguous singular cells will occur when one or more spectral classes have been omitted. The singular cell map may indicate where additional training statistics should be collected.

Using the "MAP" option causes the intermediate results tape produced by the field extraction phase of the Nonsupervised ECHO processor to replace the response value in each channel of a data vector by the average response value for that channel of the field to which the data vector belongs. A false color photo of these data channels from such an intermediate results tape may help the analyst identify areas which are spectrally homogeneous and may reduce such noise effects as the six line banding which is encountered in some Landsat data sets. Figure 2a-32 is an example object map produced by the field extraction phase of the Nonsupervised ECHO processor. Blanks on this map indicate

singular cells. Each pixel falling in an object identified by the Nonsupervised field extraction algorithm is represented by a symbol which has arbitrarily been assigned for that object.

The greatest potential aid to the training process is the intermediate tape produced by the Nonsupervised ECHO field extraction routine. This tape contains the covariance and mean matrices of each field identified (without the benefit of class statistics) by the Nonsupervised ECHO processor. This spectral/spatial cluster may contain all the information an analyst needs to produce a more representative class statistics deck for a given area in less time than would be required to produce a statistics deck by conventional methods.

Recommendations

The ECHO processors are documented in this report, the Final Technical Report for May, 1977 [5], the ECHO User's Guide [6], and an ECHO case study [7]. All the ECHO algorithms are available to JSC (and other remote terminal sites) via the LARS remote terminal. It is recommended that:

- * JSC personnel should be encouraged (or assigned) to use the ECHO algorithms via the remote terminal. Only through this type of experience will NASA personnel develop experience and confidence in using this approach to classification. Only in this way will they gain sufficient insight into the characteristics of the ECHO classifiers to appreciate the potential impact of the ECHO approach in the context of large area surveys.
- * In order to further evaluate the utility of the Nonsupervised field extraction algorithm as a method for spectral/spatial clustering, tests should be initiated involving it and ERIM's "Glob and Blob" algorithm. These tests could be carried out either by LARS or by personnel at JSC via the LARS remote terminal.
- * The general utility of the ECHO processors and training procedures and classification analyses should be investigated. For example, the singular cell map produced by the Supervised ECHO processor provides an indicator of the adequacy of the training statistics for a given area; groups of contiguous singular cells, where fields or other objects are known to exist, indicate the omission of one or more spectral classes from the statistics. Such investigation could be included in the training area selection/pixel labeling investigations. Also needed are training procedures geared specifically for the training of sample classifiers. Although these procedures have been evolving gradually, present methods are still more appropriate for training perpoint classifiers.

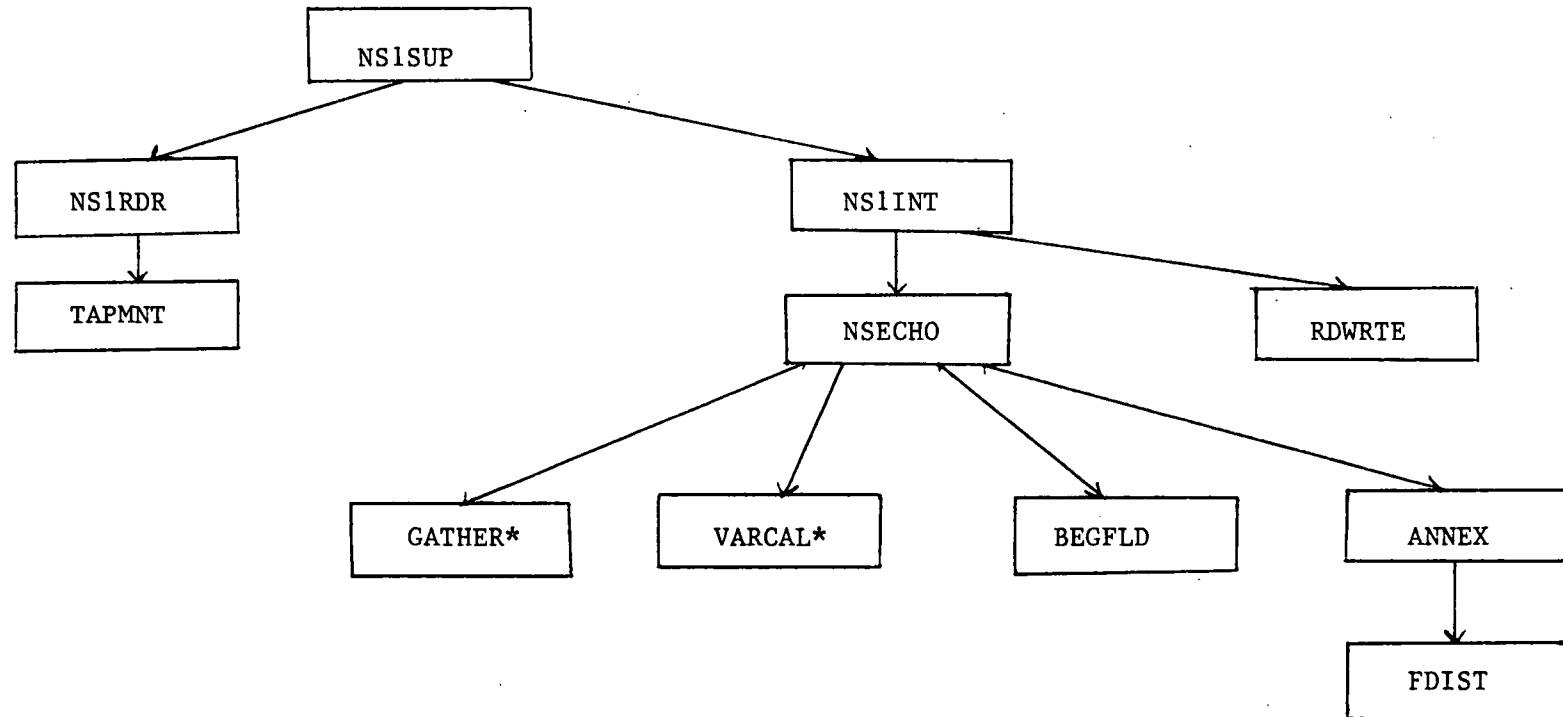
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National Aeronautics and Space Administration, Houston, Texas.
JSC 09389.

APPENDIX A

Fortran Program Documentation
for the Nonsupervised ECHO Processor

MAIN SUBROUTINES TREE OF THE FIELD EXTRACTION PHASE
OF THE NONSUPERVISED ECHO PROCESSOR



*Delivered with Supervised ECHO Documentation in the Final Technical Report on
NASA Contract NAS9-14970, May 1977

LARS Program Abstract_____

MODULE IDENTIFICATION

Module Name: NS1SUP Function Name: NS1ECHO

Purpose: Supervisor for NS1ECHO

System/Language: CMS/FORTRAN

Author: C. A. Pomalaza Date: 8/20/77

Latest Revisor: _____ Date: _____

MODULE ABSTRACT

Supervisor for the field extraction phase (phase 1) of Nonsupervised
ECHO (Extraction and Classification of Homogeneous Objects).

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1220 Potter Drive
West Lafayette, Indiana 47906

1. Module Usage

NS1SUP

CALL NS1SUP

NS1SUP is called by LARSMN to execute the field extraction phase of nonsupervised ECHO. There are no parameters involved.

2. Internal Description

NS1SUP receives control from LARSMN to perform Nonsupervised field extraction. NS1SUP calls NS1RDR to read and interpret the function control cards, then calls NS1INT to complete the initialization and compute array bases used to separate storage in the arrays for the subroutine that performs the field extraction process (NSECHO).

3. Input Description

Not applicable

4. Output Description

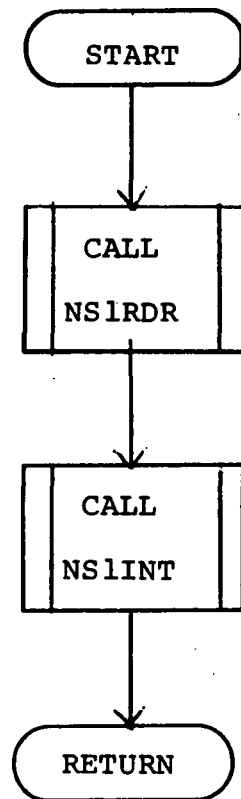
Two messages are produced and written to unit TYPEWR (the console).

UNSUPERVISED ECHO FUNCTION (PHASE 1) REQUESTED
signifies beginning of the function.

UNSUPERVISED ECHO FUNCTION (PHASE 1) COMPLETED
signifies end of the function.

5. Supplemental Information

See LARSYS System Manual for supervisor requirements.

6. Flowchart

A-5

```

C      NSISUP      LARS XXXX
C      WRITTEN      BY C. A. POMALAZA
C      *****
C      SUPERVISOR ROUTINE FOR FIELD EXTRACTION PHASE OF NONSUPERVISED ECHO
C      *****
0001      SUBROUTINE NSISUP
0002      IMPLICIT INTEGER * 4 (A-Z)
C      *****
0003      COMMON /GLCCOM/ BLANK, CARD(20), CHKOUT, COPFIL, CLASSR, CLASSX,
1      CLUSTX, CONPUT, CPYOUT, CRDRDR, CROSEQ, DATAPE,
2      DUPLTP, DUPRUN, ERRMSG, FBPNF,
3      FILESV, FLDUHD, HDATA, HEAD(88), ID(200), IMAGEX,
4      IMARK, KEYBD, MAPTAP, MAXCHA, MAXCLS,
5      PACSIZ, PCHM, POINT, PRESUX, PRNTR, READIN,
6      RSTRT, RUNFIL, RUNTAB(10,31),
7      SDATA, SEPARX, SEPTPT, SPARE(10), TEMPAS(30),
8      TPSTAT(6), TTFLOX, TYPENR,
9      TOP, ARRAY(12500)
C
0004      REAL * 8 ARRAY
0005      REAL * 4 FROCAL(5,30)
0006      INTEGER * 4 COMENT(16), DATE(5), HED1(16), HED2(16), TIME(5)
0007      INTEGER * 2 BLANK2
0008      LOGICAL * 4 CHKOUT
0009      LOGICAL * 1 BLANK1
0010      EQUIVALENCE (DATSAV, ID(1)), (CURRUN, ID(3)), (FROCAL(1), ID(51)),
1      (HED1(1), HEAD(8)), (DATE(1), HEAD(26)), (HED2(1), HEAD(39)),
2      (TIME(1), HEAD(58)), (COMENT(1), HEAD(72)),
3      (DATSAV, TPSTAT(1)),
4      (TPSCR, TPSTAT(2)), (DUPIN, TPSTAT(3)), (DASTAT, TPSTAT(4)),
5      (COPSER, TPSTAT(5)), (TRAOUT, TPSTAT(6)),
6      (BLANK, BLANK2, BLANK1)
C
0011      COMMON /NSICOM/
1      ANFLAG, ANN1, ANN2, BUFPIS, BUFRCH,
2      C1, C4, CELSZ, CELWTH,
3      CSET(3,30), CSET3(3,30), INFO(17), JPIS,
4      K9, L1, MAPFG, MAXSIZ, MINSIZ, MXSIZ, ND(200),
5      NEWRUN, NCCLS, NOFEAT, NSR, NUMCLS, NVA,
6      PIS, RESULT, ROWSIZ, ROTILE, ROTAPE, SELEC1(30),
7      TTITLE(30), TOTPTS, VARSIZ, VECISZ,
8      CSEL(30), CSEL3(30), FETVC3(30), FETVEC(30), ZDUM,
9      ARRAY4(4000)
C
0012      REAL*8 ARRAY4
0013      REAL*4 CSET, CSET3, SELEC1, ANN1, ANN2
0014      INTEGER * 2 CSEL, CSEL3, FETVEC, FETVC3
0015      LOGICAL * 4 MAPFG
0016      EQUIVALENCE (VECSIZ, NOFET3), (MXSIZ, VARSZ3), (CELWTH, GRSIZE)
C
C      CALL READER AFTER INITIAL MESSAGE PRODUCED
0017      WRITE(TYPENR, 100)
0018      100      FORMAT(' 10000 UNSUPERVISED ECHO FUNCTION (PHASE1) ',
0019      * 'REQUESTED (NSISUP)')
C      CALL NSIHD4
C
C      CALL PROCESSING ROUTINE , THEN PRINT TERMINATION MESSAGE
0020      CALL NSIINT
0021      WRITE(TYPENR, 9000)
0022      9000      FORMAT(' 10000 UNSUPERVISED ECHO FUNCTION (PHASE1) COMPLETED ',
* '(NSISUP)')

```

ECHO0010
ECHO002C
ECHO0030
ECHO0040
ECHO0050
ECHO0060
ECHO0070
ECHO0080
ECHO0090
ECHO0100
ECHO0110
ECHO0120
ECHO0130
ECHO0140
ECHO0150
ECHO0160
ECHO0170
ECHO0180
ECHO0190
ECHO0200
ECHO0210
ECHO0220
ECHO0230
ECHO0240
ECHO0250
ECHO0260
ECHO0270
ECHO0280
ECHO0290
ECHO0300
ECHO0310
ECHO0320
ECHO0330
ECHO0340
ECHO0350
ECHO0360
ECHO0370
ECHO0380
ECHO0390
ECHO0400
ECHO0410

ECHO0580
ECHO0590
ECHO0600
ECHO0610
ECHO0620
ECHO0630
ECHO0640
ECHO0650
ECHO0660
ECHO0670
ECHO0680
ECHO0690
ECHO0700
ECHO0710
ECHO0720
ECHO0730
ECHO0740
ECHO0750
ECHO0760
ECHO0770
ECHO0780

LARS Program Abstract

MODULE IDENTIFICATION

Module Name: NS1RDR

Function Name: NS1ECHO

Purpose: Reads function control cards

System/Language: CMS/FORTRAN

Author: C. A. Pomalaza

Date: 8/21/77

Latest Revisor: _____

Date: _____

MODULE ABSTRACT

NS1RDR reads and interprets all function control for the field extraction phase of Nonsupervised ECHO (Extraction and Classification of Homogeneous Objects). Checks are made for data validity. Also an intermediate results tape is readied for passing results to the classification phase of the Nonsupervised ECHO algorithm (phase 2).

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1. Module Usage

NS1RDR

CALL NS1RDR

This section lists the actions taken when the following control cards are read.

INTERMEDIATE-TAPE	The variable RQTAPE is set to the given tape number.
INTERMEDIATE-FILE	The variable RQFILE is set to the given file number.
INTERMEDIATE-INITIALIZE	The local flag INITFG is set to TRUE. This flag is used to trigger a call to TAPMNT to initialize the tape.
INTERMEDIATE-NEWRUN	A new run number is submitted to be used in the ID record of the intermediate tape.
CHANNELS	Subroutine channel is called to return VECSIZ, CSEL3, CSET3 and FETVC3 based upon interpretation of the channels card.
ANNEXATION-MEANS	Significance level 1 of unsupervised mode.
ANNEXATION-VARIANCE	Significance level 2 of unsupervised mode.
CELL-HOMOGENEITY	Cell selection (Level 1) thresholds.
CELL-WIDTH	Value of the width of a (square) cell.
PRINT-MAP	The flag MAPFLG in NS1COM is set to TRUE.

2. Internal Description

NS1RDR uses standard card reader logic in using LARSYS system sub-routines CTLWRD, CTLPRM, IVAL and FVAL in reading and interpreting the control cards. Subroutine CHANNEL is used to interpret the CHANNELS card.

NS1RDR begins by calling TSTREQ to clear the stop/suspend flag. Then flags and arrays which will convey control card information are initialized. From this point the program, functions in a loop of reading and interpreting control cards until a DATA or END card is read indicating the card of function control cards. After the control cards have been read, several checks are made on the data. A CHANNELS card must be supplied and the value of the

cell width is checked to be equal to or greater than 2. The intermediate tape is then mounted and positioned. Finally a list of selected options is printed.

3. Input Description

Function control cards for NS1ECHO are read by LARSYS system routine CTLWRD.

4. Output Description

Control card error messages are written to both the printer (PRNTR) and the console (TYPEWR). A brief list of these follows:

ERROR ON RESULTS CARD (TAPE OR FILE PARAMETER) - TYPE IN CORRECT CARD

Syntax error in the TAPE or FILE specification. Standard corrective action is taken (requesting the user retype the card from the interactive terminal).

ERROR ON CHANNELS CARD - TYPE CORRECT CARD

Standard corrective action is requested.

NO AUTO OR CHANNELS CARD - TYPE IN CHANNELS CARD

Channels must be given via the CHANNELS card. After this message the keyboard will unlock to accept a CHANNELS card. A response of carriage return will cause the keyboard to unlock again.

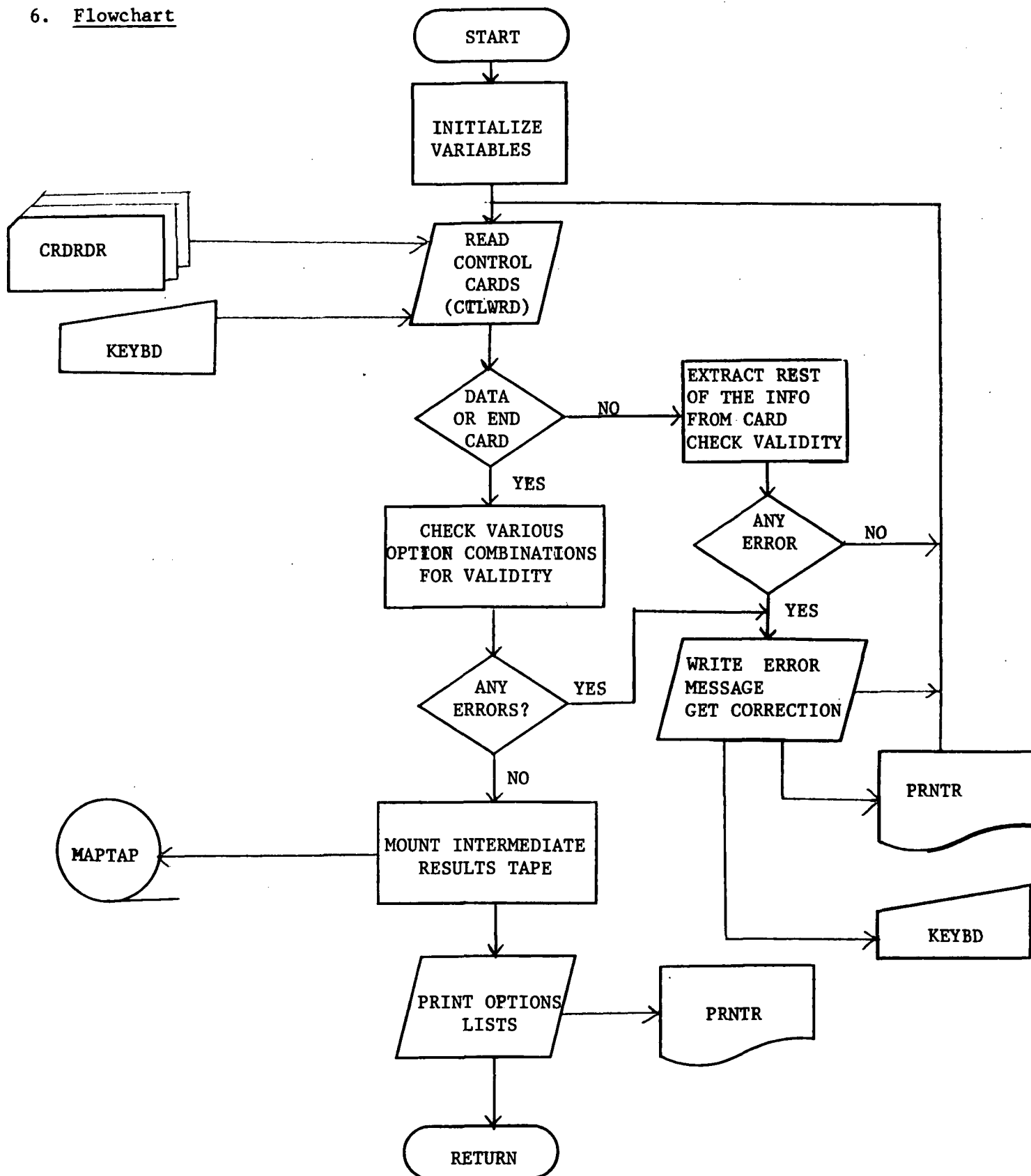
CELL SIZE MUST BE GREATER THAN OR EQUAL TO 2 - DEFAULT OF 2 ASSUMED

BOTH FILE AND INITIALIZATION OPTION REQUESTED FILE REQUESTED IGNORED.
FUNCTION CONTINUES.

The intermediate tape is initialized. Only file 1 can be initialized. In addition to this message the list of options selected is printed.

5. Supplemental Information

See LARSYS System Manual for card reader requirements.

6. Flowchart

[illegible][illegible]

```

CHANNELS CARD
.....
200 CALL CHANNELCARD, COL, VEC512, CSEL1, CSF11, FELVE1, L7401
   GO TO 101
.....
1000 ERROR CN CHANNELS CARD
.....
250 RETURN = 111
   WRITE(6,1211) VEC512
1241 FORMAT(' VEC512 IS NOW ',15)
   GO TO 110
.....
DECODE INTERMEDIATE TAPE SPECIFICATIONS
.....
300 IF(COL .EQ. 72) GO TO 101
   CALL CELPRIC(COL, COL, INTCD, 4, CODE, L111)
   IF(CODE .EQ. 4) GO TO 310
   VEC51 = 1
   CALL IVALICARD, COL, 1, VEC51, L1501
   IF(VEC51 .EQ. 0) GO TO 350
   IF(CODE .EQ. 1) NEWNUM = 1
   IF(CODE .EQ. 2) RQTAPE = 1
   IF(CODE .EQ. 3) RQFILE = 1
   GO TO 300
110 NEWNO = TRUE
   GO TO 300
.....
ERROR CN TAPE CARD
.....
500 ERRNUM = 348
   GO TO 110
.....
CELL CARD
.....
400 IF(COL .EQ. 72) GO TO 100
   CALL CELPRIC(CARD, COL, CELCOW, 2, CODE, L111)
   IF(CODE .EQ. 2) GO TO 420
   CALL IVAL(CARD, COL, 1, L1111)
   CELNUM = 1
   GO TO 400
420 IV = 10
   CALL FVALICARD, COL, AVEC, IV, L1111
   C6 = AVEC(11)
   ISEL = IV
   GO 450 11, ISEL
450 SELECT(J) = AVEC(J)
   GO TO 400
.....
DECODE ANALYTICAL CARD
.....
500 IF(COL .EQ. 72) GO TO 100
   CALL CELPRIC(CARD, COL, ANNCOD, 2, CODE, L111)
   IV = 1
   CALL FVALICARD, COL, AVEC, IV, L1111
   C6 = AVEC(11)
560 IF(CODE .EQ. 1) ANN2 = C6
   IF(CODE .EQ. 2) ANN2 = C6
   GO TO 500
.....
DECODE PRINT CARD
.....

```

[illegible]

LARS Program Abstract _____

MODULE IDENTIFICATION

Module Name: TAPMNT Function Name: Non-Supervised ECHO
Functional Support

Purpose: Mounts and positions intermediate tapes

System/Language: CMS/FORTRAN

Author: C. A. Pomalaza Date: 8/20/77

Latest Revisor: _____ Date: _____

MODULE ABSTRACT

TAPMNT mounts and positions the intermediate tape used in Phase 1 and Phase 2 of Nonsupervised ECHO.

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West Lafayette, Indiana 47906

1. Module Usage

TAPMNT

CALL TAPMNT (RQTAPE, RQFILE, MODE)

Input Arguments

RQTAPE	I*4 Number of requested tape. A tape number of 0 is a request for a scratch tape.
RQFILE	I*4 Number of requested file. If RQFILE = 0 then the tape will be initialized by writing a record type 1 (ID record) on the intermediate tape with filetype = 0.
MODE	I*4 Indicating usage of TAPMNT. MODE = 0 indicates that an intermediate results tape is being mounted with the ring out (only for reading).

Output Arguments

RQFILE	Current file position of tape.
--------	--------------------------------

2. Internal Description

See Output description.

3. Input Description

The record type of the intermediate results tape is read for each file up to and including the file needed.

4. Output Description

The following information messages are issued under the circumstances listed:

- 10042 - is typed when a tape has been mounted and before TAPMNT positions it. This message is not typed when the tape is being initialized or when the correct tape number was already mounted.
- 10043 - is typed when MODE = 1 and the file tape has results in it (Check \neq 0).
- 10045 - is typed when the tape is correctly positioned. This is not typed when initializing a tape.

After 10043 the user is asked whether he wished to overwrite the file, respecify a new intermediate results card or terminate the function.

10100 - is typed to allow entry of the new intermediate results card. This occurs when the user requests to respecify the intermediate results card.

The following error messages are typed under the conditions listed:

E361 - is written when the tape is being filed forward and a file is encountered with filetype other than zero (Check \neq 0) before the requested file is reached and MODE = 0.

E362 - is written when the circumstances for E361 occurs and MODE = 1.

For message texts refer to the User's Manual.

5. Supplemental Information

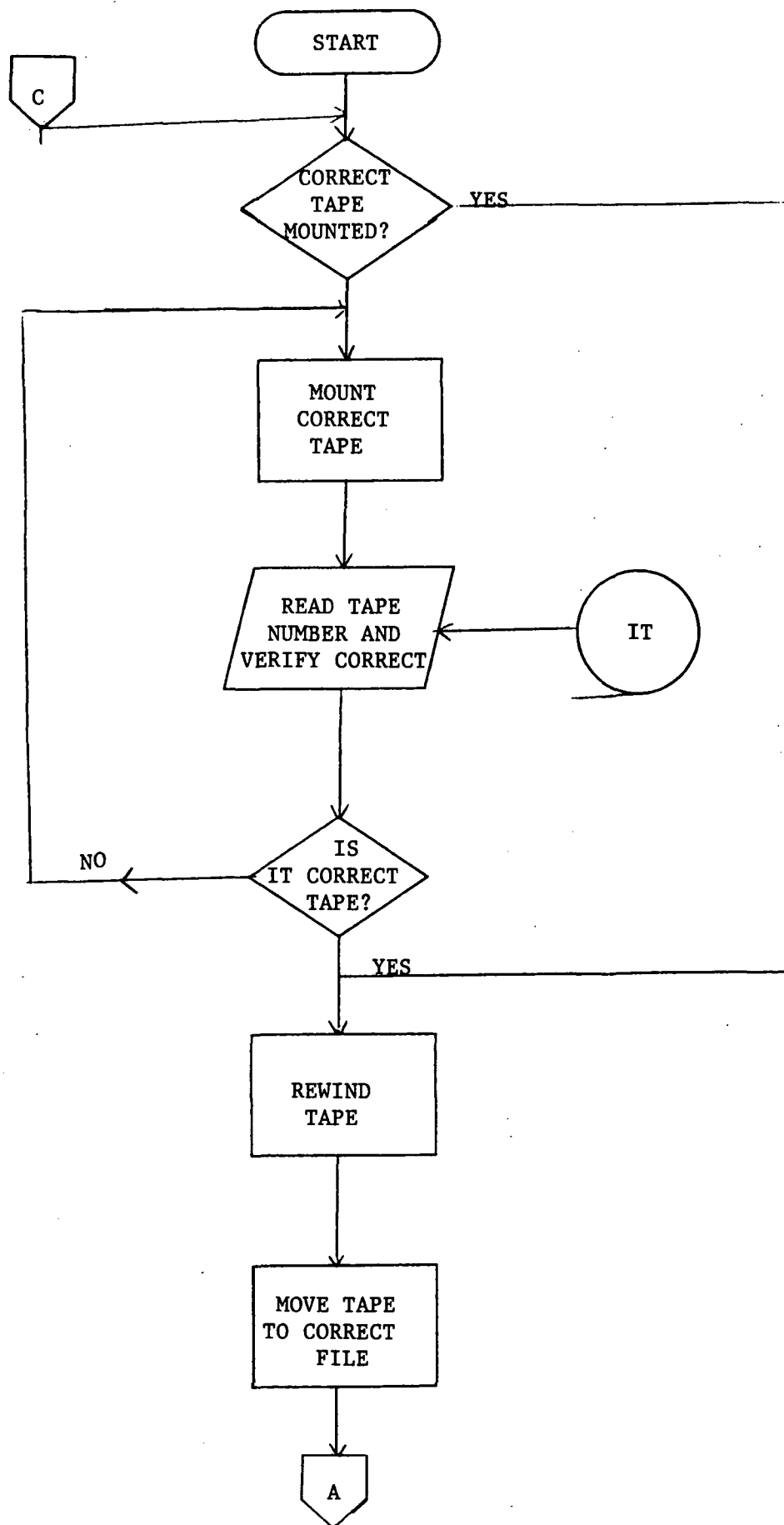
Input

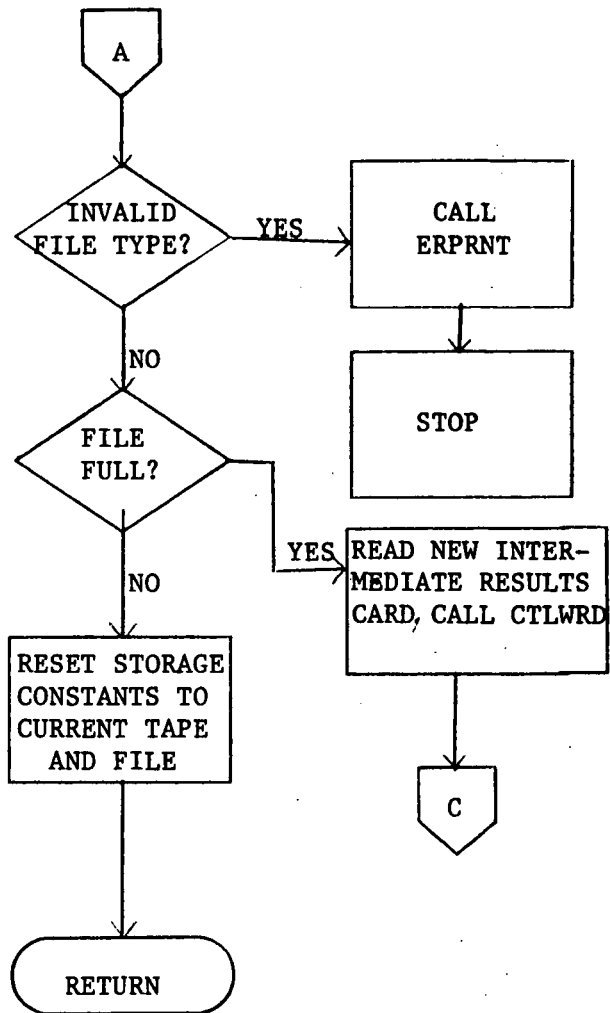
If a tape is mounted on the device and it is the incorrect tape number (as noted from the appropriate status words in CLODOM) TOPRV is called to unload the tape before the correct tape is mounted. If the correct tape is mounted TAPMNT assumes that the file number (as recorded in GLOCOM) is correct and moves the tape backwards or forwards to find the requested file.

Output

The tape is mounted with ring in for MODE = 1 and with ring out for MODE = 0. The tape is left positioned at the beginning of the requested file. When the tape is initialized LARSYS System routine TOPRW is used to do this.

6. Flowchart






```

FOXTROT IV G LEVEL 20.7          TAPMNTT          DATE = 7/23/1          22031004          PAGE 0003
FILE TAPMNT
0008      IF ICHECK .NE. 01 GO TO 230          MMT01540
0009      GO TO 400                              MMT01540
0010      CALL FPMXN(1501,MUDE,'GO')          MMT01540
0011      WRITE(UNIT,210) 1                    MMT01540
0012      WRITE(UNIT,210) 1                    MMT01540
0013      FORMAT(102,210,1)                    MMT01540
0014      CALL RTMPL                              MMT01540
0015      C                                      MMT01600
0016      230 WRITE(UNIT,220)                    MMT01600
0017      WRITE(UNIT,220)                        MMT01600
0018      430 FORMAT(1,1004) FILE HAS RESULTS IN II - DO YOU WISH TO ? MMT01600
0019      1 ' OVERWRITE THE FILE, RESPECIFY THE 'MORE RESULTS CARD, ' MMT01600
0020      ' OR TERMINATE THE JOB (OVERWRITE, RESPECIFY, TERMINATE) MMT01600
0021      ' (TAPMNT)' MMT01600
0022      235 COL = 0 MMT01600
0023      CALL CLWRP (CARD, COL, RESCD, 1, CHDL, KEYRD, ERRCD) MMT01600
0024      IF (ERRCD .EQ. 4) CALL RTMPL MMT01600
0025      GO TO 1000,240,245 MMT01600
0026      240 WRITE(UNIT,225) CODE MMT01600
0027      245 WRITE(UNIT,225) CODE MMT01600
0028      430 FORMAT(1,1010) ENTER INTERMEDIATE CARD (TAPMNT)' MMT01600
0029      CALL CLWRP (CARD, COL, INTCD, 1, CODE, KEYRD, ERRCD) MMT01600
0030      IF (ERRCD .EQ. 4) CALL RTMPL MMT01600
0031      241 IF (CDL .EQ. 2) GO TO 100 MMT01600
0032      CALL CLWRP (CARD, COL, CLACD, 1, CHDL, L245) MMT01600
0033      GO TO 1000,240,245,CODE MMT01600
0034      C                                      MMT01600
0035      244 CALL RTMPL MMT01600
0036      C                                      MMT01600
0037      245 WRITE(UNIT,225) CODE MMT01600
0038      245 WRITE(UNIT,225) CODE MMT01600
0039      C                                      MMT01600
0040      246 PROCESS PARAMETER ERRORS MMT01600
0041      246 WRITE(UNIT,225) CODE MMT01600
0042      246 FORMAT(1,24,204) MMT01600
0043      CALL EXPANT (L24,204) MMT01600
0044      GO TO 242 MMT01600
0045      C                                      MMT01600
0046      247 DECODE TAPE SPECIFICATION MMT01600
0047      250 VECSE = 1 MMT01600
0048      CALL EVAL (CARD, COL, ROTAP, VECSE, 2252) MMT01600
0049      IF (VECSZ .EQ. 0) GO TO 252 MMT01600
0050      GO TO 243 MMT01600
0051      252 L24NUM = 148 MMT01600
0052      GO TO 242 MMT01600
0053      C                                      MMT01600
0054      253 FILE SPECIFICATION MMT01600
0055      260 VECSE = 1 MMT01600
0056      CALL EVAL (CARD, COL, ROTAP, VECSE, 2252) MMT01600
0057      IF (VECSZ .EQ. 0) GO TO 252 MMT01600
0058      GO TO 243 MMT01600
0059      265 L24NUM = 147 MMT01600
0060      CALL CLWRP (CARD, COL, CLACD, 1, CHDL, L245) MMT01600
0061      GO TO 242 MMT01600
0062      C                                      MMT01600
0063      266 INITIALIZE SPECIFICATION MMT01600
0064      270 L24NUM = 1 MMT01600
0065      GO TO 243 MMT01600
0066      C                                      MMT01600
0067      271 100 WRITE(UNIT,225) CODE MMT01600
0068      430 FORMAT(1,1004) RESULTS TAPE MOUNTED AND POSITIONED'. MMT01600
0069      1 ' (TAPMNT)' MMT01600

```

```

FOXTROT IV G LEVEL 20.7          TAPMNTT          DATE = 7/23/1          22031004          PAGE 0004
FILE TAPMNT
0118      CALL TAPMNT(UNIT) MMT02240
0119      IF MUDE .EQ. 0) GO TO 400 MMT02240
0120      MAPSAV = 23TAP MMT02240
0121      FILESV = FILED MMT02240
0122      TLEUT MMT02240
0123      400 CDSLR = ROTAPE MMT02240
0124      CDPFL = FILEC MMT02240
0125      RETURN MMT02240
0126      500 CALL EXPANT(105,'STOP') MMT02240
0127      230 CALL EXPANT(201,'STOP') MMT02240
0128      END MMT02240

```

LARS Program Abstract _____

MODULE IDENTIFICATION

Module Name: NS1INT Function Name: NS1ECHO

Purpose: Initialization Including array base computation

System/Language: CMS/FORTRAN

Author: C. A. Pomalaza Date: 8/23/77

Latest Revisor: _____ Date: _____

MODULE ABSTRACT

NS1INT reads the area to be processed and computes the array bases for storage allocation of the variables used by the subroutine that performs the field extraction phase (NSECHO).

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1. Module Usage

NS1INT

CALL NS1INT

There are no arguments passed to NS1INT. It is called by NS1SUP to allocate storage prior to the call of NSECHO.

2. Internal Description

NS1INT performs the following functions:

- 1 - The data card of the area to be processed is read and interpreted using LARSYS system routine LAREAD.
- 2 - The new ID record to be used in the intermediate results tape is built using part of the ID record from the multispectral image storage tape, (MIST) and information read by NS1RDR.
- 3 - Check the validity of the channels requested comparing them with the ones present on the MIST.
- 4 - Compute the space needed for arrays to be used.
- 5 - Prints data summary.
- 6 - Call NSECHO to perform the field extraction phase.
- 7 - Call RDWRTE which transfer results written on disk by NSECHO to the intermediate results tape.

3. Input Description

The field description card of the area to be processed is read by a call to LARSYS system routine LAREAD.

4. Output Description

Information concerning the parameters and the field selected are written on the printer (PRNTR).

Information and error messages are also printed.

LAST SAMPLE NUMBER XXXXX OF FIELD XXXXXXXXX EXCEEDS LAST SAMPLE ON TAPE.
LAST SAMPLE RESET TO XXXXX.

The last sample number is reset to the one present on the MIST tape.

YOU HAVE REQUESTED A CHANNEL NOT AVAILABLE IN THIS RUN. REQUEST CANCELLED.

A channel number was requested which does not exist on the data run.

FIELDS EXCEEDS LIMITS OF DATA. FIELD IGNORED. FIELD DESIGNATION FOLLOWS:

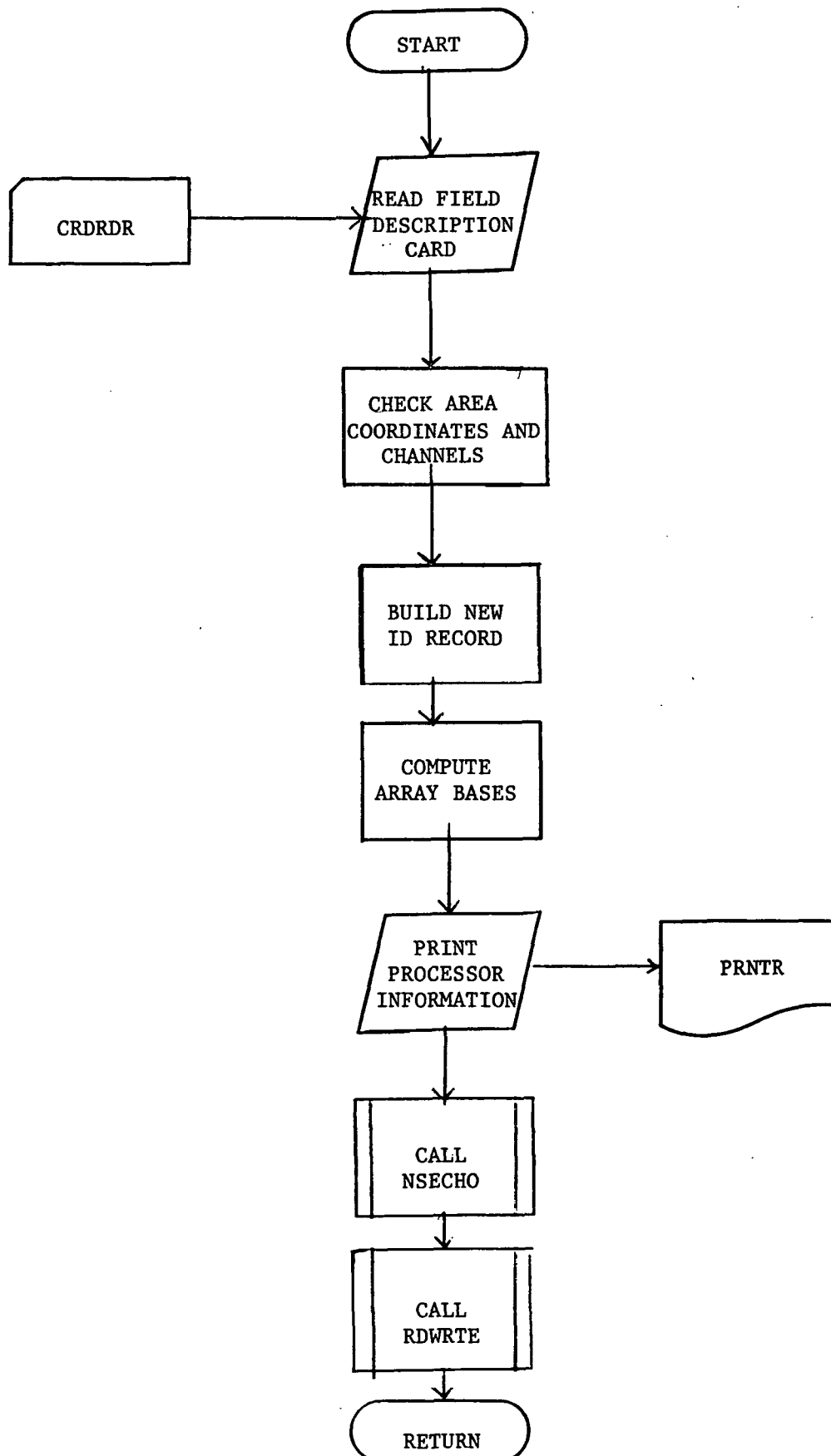
The first column of the field is greater than the last column of data on the tape.

A DATA DECK IS MISSING - FUNCTION TERMINATED

The user has forgotten to include the field description card.

5. Supplemental Information

Not applicable.

6. Flowchart

```

C#00001 0
C#00002 0
C#00003 0
C#00004 0
C#00005 0
C#00006 0
C#00007 0
C#00008 0
C#00009 0
C#00010 0
C#00011 0
C#00012 0
C#00013 0
C#00014 0
C#00015 0
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```

[illegible]

```
CMD0154C
**CMD0154D
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**CMD01570
**CMD01580
**CMD01590
**CMD01600
**CMD01610
**CMD01620
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**CMD02150
**CMD02160
**CMD02170
**CMD02180
**CMD02190
**CMD02200C
**CMD02200D
**CMD02210
**CMD02220
**CMD02230
**CMD02240
**CMD02250
**CMD02260
**CMD02270
**CMD02280
```

[illegible]

*ECH03050
*ECH03060
ECH03070
ECH03080

LARS Program Abstract _____

MODULE IDENTIFICATION

Module Name: NSECHO Function Name: NSIECHO

Purpose: Perform field extraction and disk filing for NSIECHO

System/Language: CMS/FORTRAN

Author: _____ Date: _____

Latest Revisor: C. A. Pomalaza Date: 8/21/77

MODULE ABSTRACT

NSECHO computes cell mean and variances row by row, performs cell splitting, annexation and field closures. As fields are annexed this information is sequentially stored on disk. As fields are closed, field statistics information is stored in a random access file on disk.

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1. Module UsageNSECHO

CALL NSECHO (STACK, CLOSED, OPEN, FLDSIZ, PRTBUF, RDATAT, PIXCOR, PIXVAL, CELSUM, CELCOR, FLDSUM, FLDCOR, CELVAR, FLDVAR, CELAVE, AUXSUM, AUXVAR, WORK, BUFFER, LINBUF, BDATA, RDATA, TAPBUF)

All arguments are array bases computed by NSIINT.

STACK	I*4 Array dimensioned JPTS, where JPTS is the number of cell widths/line being processed. It is a revolving queue to store flags about the field status (which field storage areas are in use).
CLOSED OPEN	L*1 Arrays each dimensioned JPTS. Logical flags for each field, used to ascertain closure status.
FLDSIZ	I*4 Array dimensioned JPTS. Keeps record of number of pixels in each field.
PRTBUF	I*2 Array used to store output symbols to print object map (dimensioned by core constraints).
PIXVAL	R*4 Array dimensioned VECSIZ * CELSIZ where VECSIZ is the number of channels used and CELSIZ is the number of pixels/cell used to store cell values during processing.
PIXCOR	R*4 Dimensioned $\text{VECSIZ} * \text{MTXSIZ}$ where $\text{MTXSIZ} = \frac{\text{VECSIZ} * (\text{VECSIZ} + 1)}{2}$. Used to store intermediate correlation matrices.
CELSUM	R*4 Array dimensioned $\text{VECSIZ} * \text{JPTS}$ and used to store cell means one row at a time.
CELCOR	R*4 Array dimensioned $\text{MTXSIZ} * \text{JPTS}$ and used to store cell correlation matrices one row at a time.
FLDSUM	R*4 Array dimensioned $\text{VECSIZ} * \text{JPTS}$ used to accumulate field means.
FLDCOR	R*4 Array dimensioned $\text{MTXSIZ} * \text{JPTS}$ used to accumulate field correlation.
CELVAR FLDVAR	R*4 Arrays dimensioned $\text{JPTS} * \text{VECSIZ}$ and used to store cell and field variances.
CELAVE	R*4 Array dimensioned VECSIZ used as intermediate cell mean buffer.

AUXCOR	R*4 Arrays dimensioned JPTS * VECSIZ used as auxiliary storage during variance processing for candidate annexations.
AUXSUM	R*4 Array dimensioned JPTS * VECSIZ used to store means for candidate annexations.
WORK	R*4 Working area. What remains of ARRAY in GLOCOM after ZCOV.
BUFFER	I*2 Hold field assignments from current and previous line of data. Dimensioned 2*JPTS*2. Buffer (1,x) contains the field number of cell x, BUFFER (2,x) contains the relative field pointer into FLDCOV (covariance of open fields).
BDATA	L*1 Array dimensioned VECSIZ * ID(6). Receives raw data values.
RDATA	R*4 Dimensioned VECSIZ * JPTS * CELSIZ holds calibrated data values from the tape.
TAPBUF	I*4 Symbol storage area.
LINBUF	I*4 Output storage area.

2. Internal Description

NSECHO operates in a loop over the lines requested. It retrieves a row of data, annexes cells into fields on a column basis. It then writes the buffer line to disk. Reads the next row (cell width lines) annexes it by columns and by rows to the first row. Fields that are closed are written with the field number to the intermediate tape and the buffer is written to disk. The slots in FLDCOV are then released and a new row is read. This processing continues until the lines are completed. NSECHO uses GATHER to compute cell variances and means, ANNEX to perform annexation and BEGFLD to open new fields.

3. Input Description

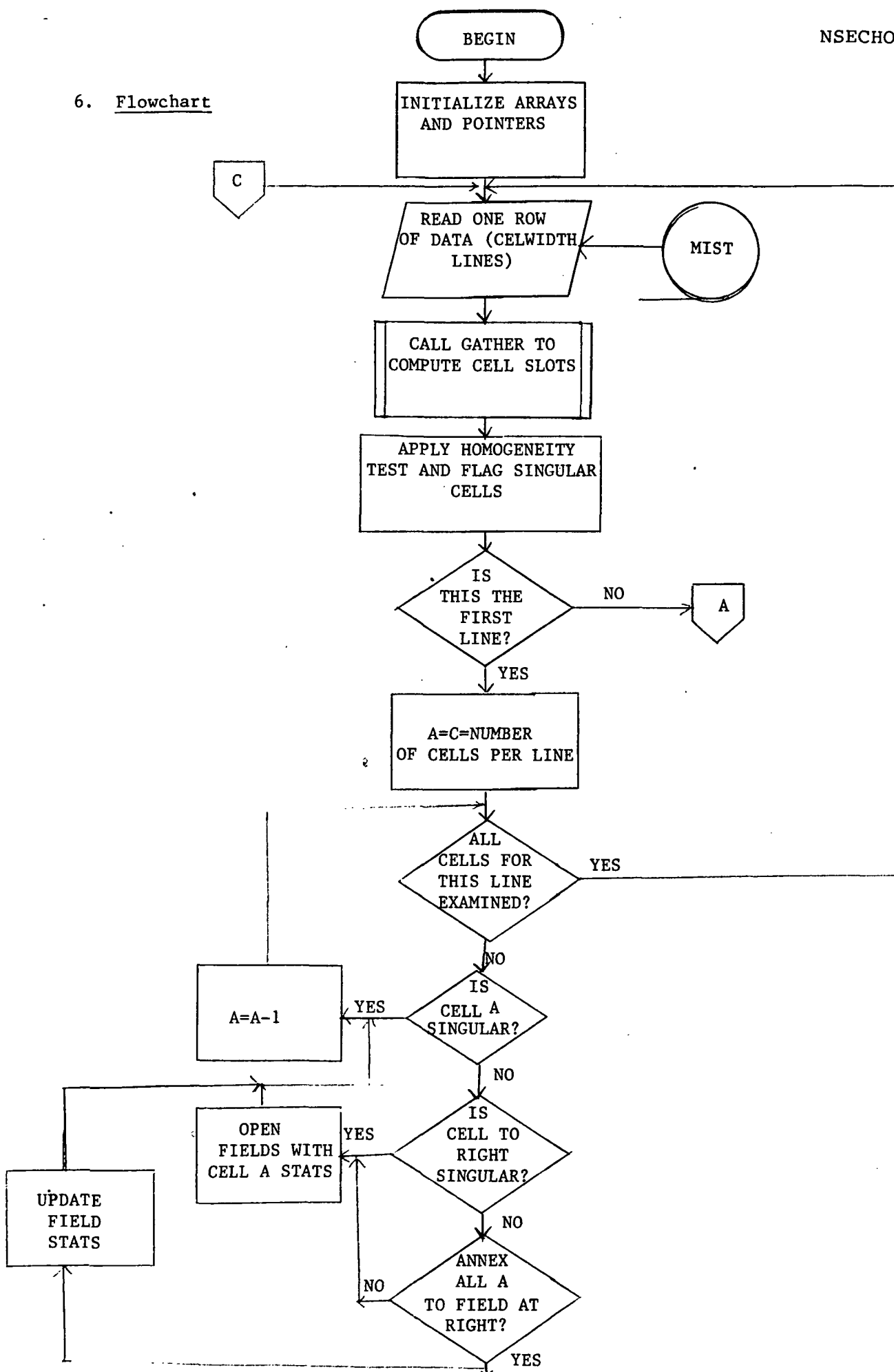
Data is read from tape via GADLIN.

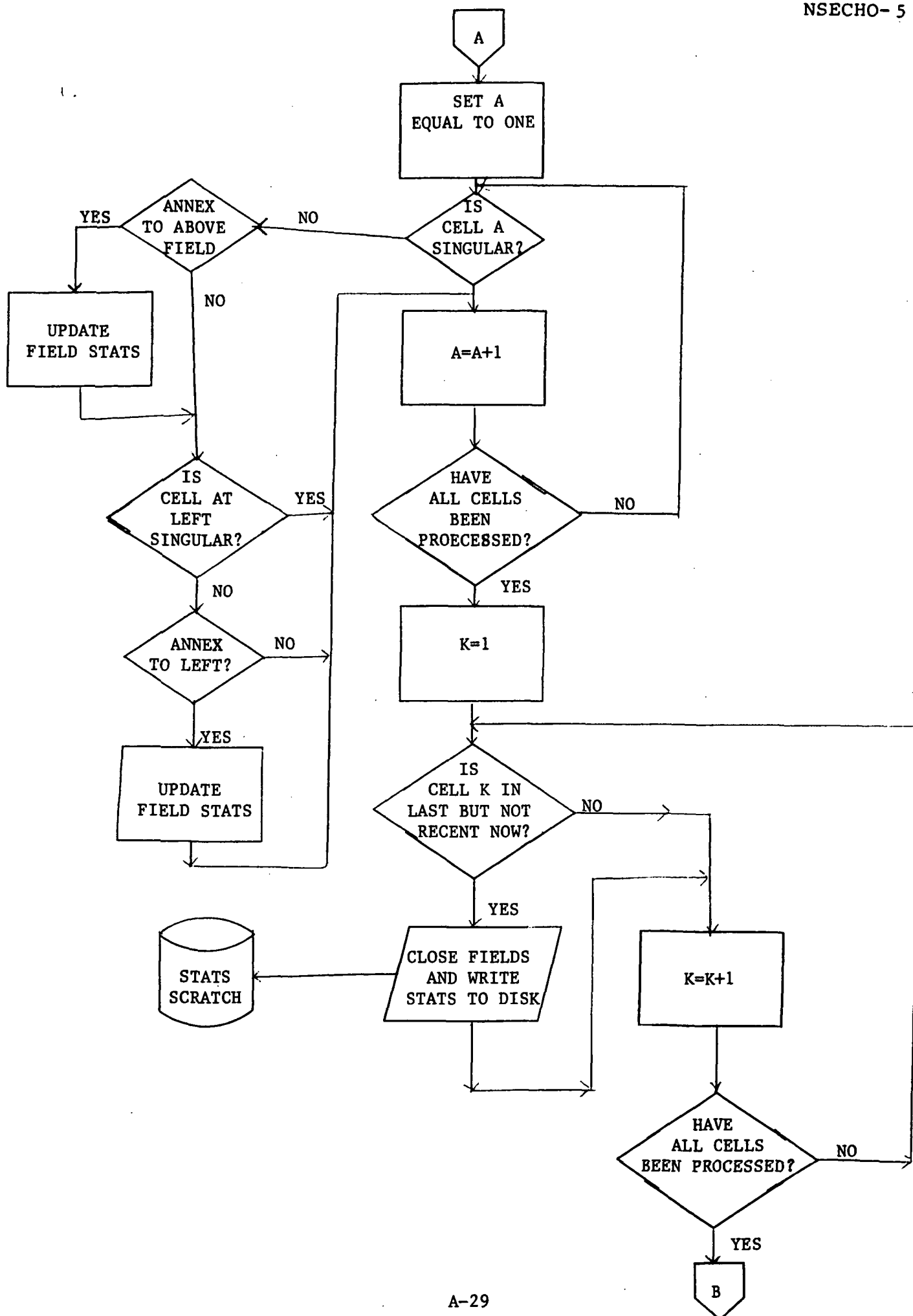
4. Output Description

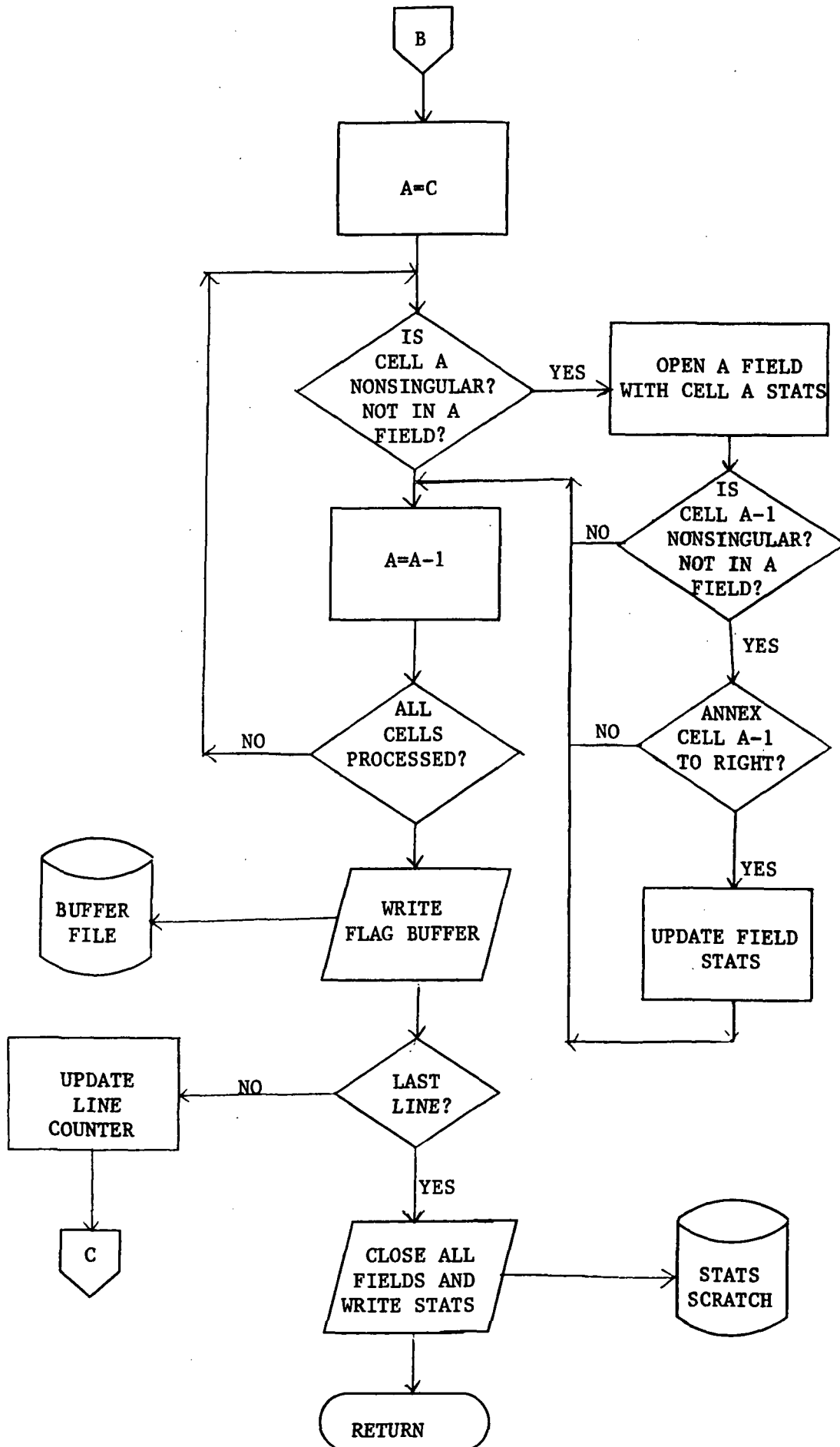
Information is written on disk using unformatted write.

5. Supplemental Information

See BUFFER and SCRATCH file description.

6. Flowchart





1 CH000110
 1 CH000120
 1 CH000130
 1 CH000140
 1 CH000150
 1 CH000160
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 1 CH000200
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PAGE 0002

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```

C .....
C INITIALIZE .....
C .....
0072 JPTS = JPTS. ECHO1550
0073 BOTTOM = JPTS. ECHO1560
0074 DO 100 I = 2,4 ECHO1570
0075 100 BLOCK(I) = INFO(I)+51 ECHO1580
0076 DO 110 J = 1,JPTS ECHO1590
0077 STACK(I) = 1 ECHO1600
0078 PRIBUF(I) = BLANK2 ECHO1610
0079 CLCSED(I) = .TRUE. ECHO1620
0080 110 OPEN(I) = .FALSE. ECHO1630
0081 CLC = LIN*CELWTH ECHO1640
0082 C1 = CELSIZ/2 ECHO1650
0083 LETCON = SLODATE*(VECSIZ) ECHO1660
0084 R0 = (VECSIZ-1)*VECSIZ/2 ECHO1670
0085 R7 = (VECSIZ-1)*(VECSIZ+1)/6 ECHO1680
0086 DO 120 J = 1,VECSIZ ECHO1690
0087 R4(J) = SELEC(I) ECHO1700
0088 120 CONTINUE ECHO1710
0089 NOLINE = 0 ECHO1720
0090 PLIN = FSTLIN ECHO1730
0091 NLT = 1 ECHO1740
0092 ISYM = 0 ECHO1750
0093 PRIRW = 1 ECHO1760
0094 JPT = JPTS - 1 ECHO1770
0095 BLDW = 0 ECHO1780
0096 NEWRW = 1 ECHO1790
0097 CURLIN = FSTLIN ECHO1800
0098 C THIS IS THE BEGINNING OF THE OLD SUB LOOP ECHO1810
0099 150 NOLINE = NOLINE + 1 ECHO1820
0100 I = CURLIN ECHO1830
0101 DO 155 J = 1,CELWTH ECHO1840
0102 BLOCK(I) = 1 ECHO1850
0103 CALL GADLIN(BLOCK(I),CSEL3,CSET3,ID,DATE,VECSIZ,NSR,BDATE, ECHO1860
0104 & RDATA(I),JPT,ERROR) ECHO1870
0105 IF (ERROR.GT.0) CALL LINERRERRR(1,615,615,615) ECHO1880
0106 155 I = I + 1 ECHO1890
0107 160 M = 1 ECHO1900
0108 CALL GATHER(DATA(I),JPT,PIXVAL(PIXCOR,CELSUM(I,CELSUM), ECHO1910
0109 & CELCOR(CELSUM),CELSUM,VECSIZ,MXSIZ,NSR,NVA) ECHO1920
0110 201 M = M + 1 ECHO1930
C .....
C ENTER MAIN LOOP -- READ CELL SUMS AND CORRELATIONS. ECHO1940
C .....
0111 151 J = NOLINE*CELSUM ECHO1950
0112 IF (INDIDLINE.EQ.0) WRITE(1,PEWR,9235) J ECHO1960
0113 9235 FORMAT(16,' LINES CLASSIFIED: (ECHO1970
0114 & ROWS = 1 + (CURLIN-1)*(JPTS+4) ECHO1980
0115 ROWS = 1 + (NEWRW-1)*(JPTS+4) ECHO1990
0116 PROBS = (NEWRW-JPTS+4) ECHO2000
C .....
C GET ONE ROW OF DATA. ECHO2010
C .....
C COMPUTE CELL STATISTICS AND SCREEN OUT BAD CELLS. ECHO2020
C .....
0117 J = NRCWBS ECHO2030
0118 I2 = 1 ECHO2040
0119 I3 = 1 ECHO2050
0120 I4 = 1 ECHO2060
0121 I5 = 1 ECHO2070
0122 I6 = 1 ECHO2080
0123 I7 = 1 ECHO2090
0124 ECHO2100

```

```

0125 DO 200 I = 1,JPTS ECHO2110
0126 BUFFER(I) = 0 ECHO2120
C .....
C RETURN VARIANCES FROM SUM AND CROSS PRODUCT MTR ECHO2130
C .....
0127 170 CALL VARGAL(CELSUM(I,12),CELAVE,CELCOR(I,12), ECHO2140
0128 & CELVAR(I,15),CELSIZ,VECSIZ) ECHO2150
0129 176 DO 177 K = 1,VECSIZ ECHO2160
0130 IF (CELVAR(K,15).GT. R9(K)) CELSUM(K,12)*CELAVE(K) GO TO 190 ECHO2170
0131 177 CONTINUE ECHO2180
0132 GO TO 198 ECHO2190
C .....
C INHOMOGENEOUS CELL. CLASSIFY IT SEPARATELY. ECHO2200
C .....
0132 190 CONTINUE ECHO2210
0133 K = J + 1 ECHO2220
0134 DO 196 L = J,K ECHO2230
0135 196 BUFFER(L) = 1 ECHO2240
0136 PRIBUF(I) = SING ECHO2250
0137 I2 = I2 + 1 ECHO2260
0138 I3 = I3 + 1 ECHO2270
0139 I4 = I4 + 1 ECHO2280
0140 I5 = I5 + 1 ECHO2290
0141 IF (ULDRW.EQ.0) GO TO 265 ECHO2300
C .....
C SCAN ULDRW FOR VERTICAL BOUNDARIES AND SINGULARITIES. ECHO2310
C .....
0141 K = ROWWBS ECHO2320
0142 GO TO 228 ECHO2330
C .....
C COMPARE CELLS WITH FIELDS. ECHO2340
C .....
C COMPARE CURRENT ROW WITH PREVIOUS ROW. ECHO2350
C .....
0143 J = ROWWBS ECHO2360
0144 K = ROWWBS ECHO2370
0145 DO 240 I = 1,JPTS ECHO2380
0146 M = BUFFER(I) ECHO2390
0147 N = BUFFER(I+1) ECHO2400
0148 IF (M.EQ.0) AND (N.EQ.0) CLCSED(I) = .FALSE. ECHO2410
0149 IF (BUFFER(K) = 0) GO TO 239 ECHO2420
0150 I5 = I5 + 1 ECHO2430
0151 239 IF (FLAG) 230,230,227 ECHO2440
0152 227 CPEV(I) = .TRUE. ECHO2450
0153 BUFFER(K) = M ECHO2460
0154 BUFFER(K+1) = N ECHO2470
0155 PRIBUF(I) = OLBUFF(I) ECHO2480
0156 GO TO 239 ECHO2490
C .....
C COMPARE WITH FIELD TO LEFT, IF ANY. ECHO2500
C .....
0159 230 IF (I.EQ.1) GO TO 239 ECHO2510
0160 M = BUFFER(I-1) ECHO2520
0161 N = BUFFER(I-2) ECHO2530
0162 IF (M.EQ.0) GO TO 239 ECHO2540
0163 239 230 GO TO 239 ECHO2550
0164 232 IF (FLAG) 2305,2305,238 ECHO2560
0165 235 CPEV(I) = .TRUE. ECHO2570
0166 238 BUFFER(K) = M ECHO2580
0167 238 BUFFER(K+1) = N ECHO2590
0168 PRIBUF(I) = PRIBUF(I-1) ECHO2600
0169 GO TO 239 ECHO2610
0170 ECHO2620

```

```

FILE NSECHOL
0171 2845 ISYM = 1 + MOD(1SYM,16) ECHO3070
0172 2846 JPT5 = 5 + 4 ECHO3080
0173 2847 K = K + 4 ECHO3090
0174 2848 K = K + 4 ECHO3100
C ***** ECHO3110
C ***** ECHO3120
C ***** ECHO3130
C ***** ECHO3140
C ***** ECHO3150
C ***** ECHO3160
C ***** ECHO3170
C ***** ECHO3180
C ***** ECHO3190
C ***** ECHO3200
C ***** ECHO3210
C ***** ECHO3220
C ***** ECHO3230
C ***** ECHO3240
C ***** ECHO3250
C ***** ECHO3260
C ***** ECHO3270
C ***** ECHO3280
C ***** ECHO3290
C ***** ECHO3300
C ***** ECHO3310
C ***** ECHO3320
C ***** ECHO3330
C ***** ECHO3340
C ***** ECHO3350
C ***** ECHO3360
C ***** ECHO3370
C ***** ECHO3380
C ***** ECHO3390
C ***** ECHO3400
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C ***** ECHO3440
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C ***** ECHO3460
C ***** ECHO3470
C ***** ECHO3480
C ***** ECHO3490
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C ***** ECHO3570
C ***** ECHO3580
C ***** ECHO3590
C ***** ECHO3600
C ***** ECHO3610
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C ***** ECHO3670
C ***** ECHO3680
C ***** ECHO3690
C ***** ECHO3700
C ***** ECHO3710
C ***** ECHO3720
C ***** ECHO3730
C ***** ECHO3740
C ***** ECHO3750
C ***** ECHO3760
C ***** ECHO3770
C ***** ECHO3780
C ***** ECHO3790
C ***** ECHO3800
C ***** ECHO3810
C ***** ECHO3820

```

```

FILE NSECHOL
0221 5240 FORMAT(1X,11111) ECHO3830
0222 5241 JK = 1 + JPT5 ECHO3840
0223 5242 CLOUP(JK) = PRTOUF(JK) ECHO3850
0224 5243 CLOUP(JK) = CLOUP(JK) ECHO3860
C ***** ECHO3870
C ***** ECHO3880
C ***** ECHO3890
C ***** ECHO3900
C ***** ECHO3910
C ***** ECHO3920
C ***** ECHO3930
C ***** ECHO3940
C ***** ECHO3950
C ***** ECHO3960
C ***** ECHO3970
C ***** ECHO3980
C ***** ECHO3990
C ***** ECHO4000
C ***** ECHO4010
C ***** ECHO4020
C ***** ECHO4030
C ***** ECHO4040
C ***** ECHO4050
C ***** ECHO4060
C ***** ECHO4070
C ***** ECHO4080
C ***** ECHO4090
C ***** ECHO4100
C ***** ECHO4110
C ***** ECHO4120
C ***** ECHO4130
C ***** ECHO4140
C ***** ECHO4150
C ***** ECHO4160
C ***** ECHO4170
C ***** ECHO4180
C ***** ECHO4190
C ***** ECHO4200
C ***** ECHO4210
C ***** ECHO4220
C ***** ECHO4230
C ***** ECHO4240
C ***** ECHO4250
C ***** ECHO4260
C ***** ECHO4270
C ***** ECHO4280
C ***** ECHO4290
C ***** ECHO4300
C ***** ECHO4310
C ***** ECHO4320
C ***** ECHO4330
C ***** ECHO4340
C ***** ECHO4350
C ***** ECHO4360
C ***** ECHO4370
C ***** ECHO4380
C ***** ECHO4390
C ***** ECHO4400
C ***** ECHO4410
C ***** ECHO4420
C ***** ECHO4430
C ***** ECHO4440
C ***** ECHO4450
C ***** ECHO4460
C ***** ECHO4470
C ***** ECHO4480
C ***** ECHO4490
C ***** ECHO4500
C ***** ECHO4510
C ***** ECHO4520
C ***** ECHO4530
C ***** ECHO4540
C ***** ECHO4550
C ***** ECHO4560
C ***** ECHO4570
C ***** ECHO4580

```

```

FILE NSECHOL
0260 5260 IF = 16 + CELWTH - 1 ECHO4590
0261 5261 CC 4001 16,16,17 ECHO4600
0262 5262 IS = 15 + 1 ECHO4610
0263 5263 CC 17,4002 ECHO4620
0264 5264 PRTOUF(12) = BUFFER(18) - JPT5 ECHO4630
0265 5265 PRTOUF(12) = PRTOUF(12) ECHO4640
0266 5266 PRTOUF(12) = PRTOUF(12) ECHO4650
0267 5267 CC 17,190 ECHO4660
0268 5268 END ECHO4670

```

LARS Program Abstract _____

MODULE IDENTIFICATION

Module Name: BEGFLD Function Name: NS1ECHO

Purpose: Initialization of field statistics

System/Language: CMS/FORTRAN

Author: _____ Date: _____

Latest Revisor: C. A. Pomalaza Date: 8/21/77

MODULE ABSTRACT

BEGFLD initializes field statistics from cell statistics for the field extraction phase of unsupervised ECHO.

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West Lafayette, Indiana 47906

1. Module UsageBEGFLD

CALL BEGFLD (CELSUM, FLDSUM, CELCOR, FLDCOR, CELVAR, FLDVAR, FLDSIZ)

Input Arguments

CELSUM	R*4 Array dimensioned VECSIZ and used to store cell means.
CELCOR	R*4 Array dimensioned MTXSIZ and used to store cell correlation matrix.
CELVAR	R*4 Array dimensioned VECSIZ and used to store cell variances.

Output Arguments

FLDSUM	R*4 Array dimensioned VECSIZ and used to store field means.
FLDCOR	R*4 Array dimensioned MTXSIZ and used to store field correlation matrix.
FLDVAR	R*4 Array dimensioned VECSIZ and used to store field variances.
FLDSIZ	I*4 Variable to keep record of the number of pixels in each field.

2. Internal Description

BEGFLD initializes FLDSUM, FLDCOR, FLDVAR and FLDSIZ to be equal to CELSUM, CELCOR, CELVAR, and CELSIZ.

3. Input Description

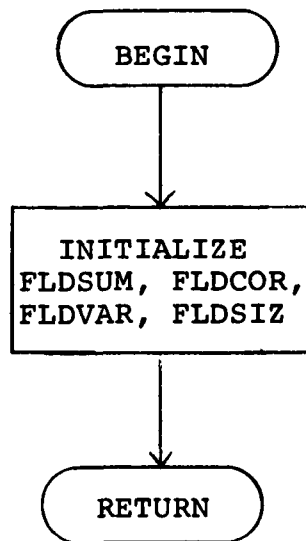
Not applicable.

4. Output Description

Not applicable.

5. Supplemental Information

Not applicable.

6. Flowchart

FILE BEGFLO

```

C BEGFLO          MODIFIED BY C.A. POMALAZA          EC400010
C INITIALIZES STATISTICS FOR A NEW FIELD          EC400020
0001      SUBROUTINE BEGFLO(CELSUM,FLOSUM,CELCOR,FLOCOR,CELVAR,FLOVAR,          EC400030
          & FLODSIZ)          EC400040
0002      IMPLICIT INTEGER * 4 (A-Z)          EC400050
C-----          EC400060
0003      COMMON /NSICOM/          EC400070
          1 ANFLAG, ANN1, ANN2, BUFPTS, BUFROZ,
          2 C1, C4, CELSIZ, CELWTH,
          3 CSET(3,30), CSET3(3,30), INFO(17), JPTS,
          4 K9, L1, MAPFLG, MAXSIZ, MINSIZ, MTXSIZ, NC(200),
          5 NEWRUN, NOCLS, NOFEAT, NSR, NUMCLS, NVR,
          6 PTS, RESULT, ROWSIZ, RQFILE, RQTAPE, SELEC1(30),
          7 TTITLE(30), TOTPTS, VARSIZ, VECSIZ,
          8 CSEL(30), CSEL3(30), FETVC3(30), FETVEC(30), ZDUM,
          9 ARRAY4(4000)
0004      REAL*8 ARRAY4
0005      REAL*4 CSET,CSET3,SELEC1,ANN1,ANN2
0006      INTEGER * 2 CSEL,CSEL3,FETVEC,FETVC3          EC400240
0007      LOGICAL * 4 MAPFLG          EC400250
0008      EQUIVALENCE (VECSIZ,NOFET3),(MTXSIZ,VARSZ3),(CELWTH,GRSIZE)          EC400260
0009      REAL * 4 CELSUM(VECSIZ),FLOSUM(VECSIZ),CELCOR(MTXSIZ),          EC400270
          & FLOCOR(MTXSIZ),CELVAR(VECSIZ),FLOVAR(VECSIZ)          EC400280
C-----          EC400290
0010      FLODSIZ = CELSIZ          EC400300
0011      DO 1 I=1,VECSIZ          EC400310
0012      FLOVAR(I) = CELVAR(I)          EC400320
0013      FLOSUM(I) = CELSUM(I)          EC400330
0014      DO 2 I=1,MTXSIZ          EC400340
0015      FLOCOR(I) = CELCOR(I)          EC400350
0016      RETURN          EC400360
0017      END          EC400370
          EC400380
          EC400390
          EC400400
          EC400410

```


LARS Program Abstract _____

MODULE IDENTIFICATION

Module Name: ANNEX Function Name: NS1ECHO

Purpose: Carry out annexation tests and processing

System/Language: CMS/FORTRAN

Author: _____ Date: _____

Latest Revisor: C. A. Pomalaza Date: 8/21/77

MODULE ABSTRACT

ANNEX performs the multiple-univariate tests for annexation of cells to fields and updates field statistics when the annexation criteria are achieved.

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1. Module UsageANNEX

CALL ANNEX (CELSUM, FLDSUM, AUXSUM, CELCOR, FLDCOR, CELVAR, FLDVAR, AUXVAR, FLDSIZ)

Input Argument

CELSUM	R*4 Arrays dimensioned VECSIZ and used to store cell and field means respectively.
FLDSUM	
CELCOR	R*4 Arrays dimensioned MTXSIZ and used to store cell and field correlation matrices.
FLDCOR	
CELVAR	R*4 Arrays dimensioned VECSIZ and used to store cell and field variances.
FLDVAR	
FLDSIZ	I*4 Variable that records the number of pixels in each field.

Output Arguments

FLDSUM	R*4 See Above.
AUXSUM	R*4 Array dimensioned VECSIZ and used to store means for candidate field annexations.
FLDCOR	R*4 Array. See above.
FLDVAR	R*4 Array. See above.
AUXVAR	R*4 Array dimensioned VECSIZ and used to store variances for candidate annexations.

2. Internal Descriptions

ANNEX performs the multiple-univariate tests for equivalent mean vectors and for equivalent covariance matrices between a cell and a field. If both tests are successful the cell is annexed to the field and the field statistics are updated. ANNEX calls the function FDIST to find the decision threshold values of the F-distribution necessary to perform the tests.

3. Input Description

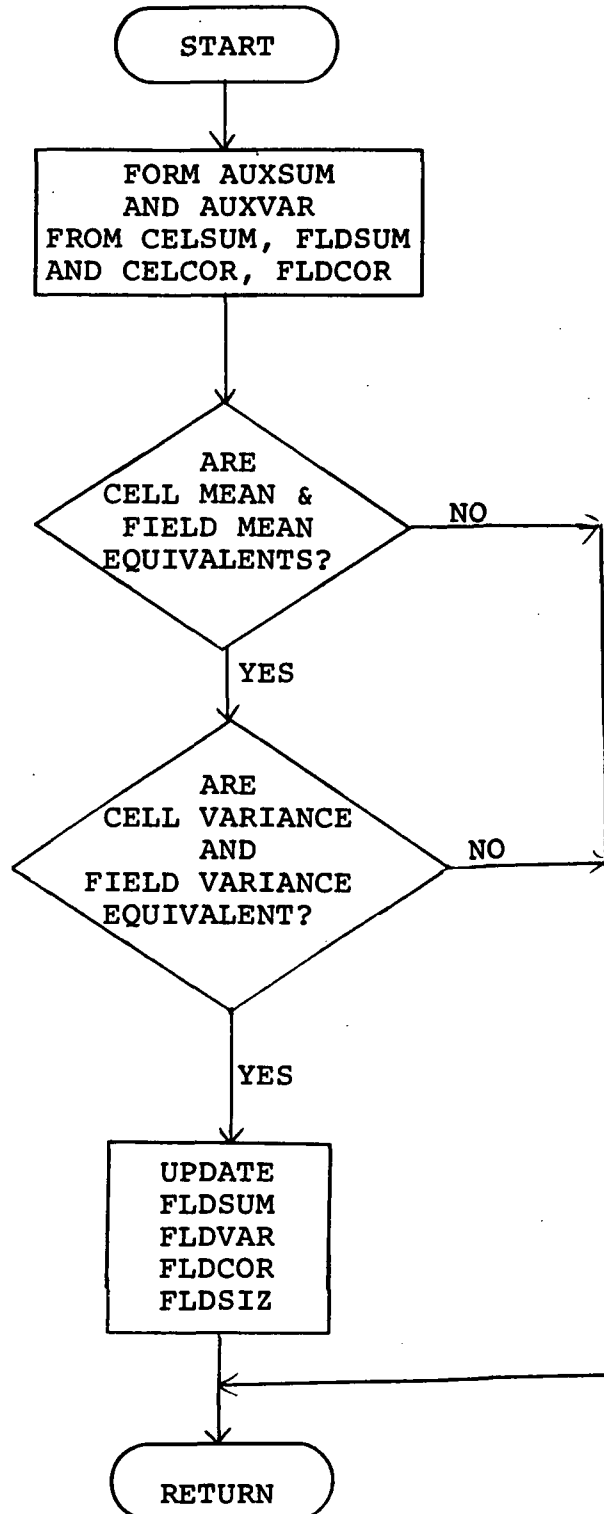
Not applicable

4. Output Description

Not applicable.

5. Supplemental Information

Not applicable.

6. Flowchart

FILE ANNEX

```

C
C ANNEX      MODIFIED BY C.A. POMALAZA
C
0001      SUBROUTINE ANNEX(CELSUM,FLOSUM,AUXSUM,CELCOR,FLOCOR,CELVAR,
0002      C      FLOVAR,AUXVAR,FLOSIZ)
C      IMPLICIT INTEGER * 4 (A-Z)
C
0003      COMMON /NSICOM/
C      1 ANFLAG, ANN1, ANN2, BUFPPTS, BUFRQZ,
C      2 C1, C4, CELSIZ, CELWTH,
C      3 CSET(3,30), CSET3(3,30), INFO(17), JPTS,
C      4 K9, LI, MAPFLG, MAXSIZ, MINSIZ, MTXSIZ, ND(200),
C      5 NEWRUN, NCCLS, NOFEAT, NSR, NUMCLS, NVR,
C      6 PTS, RESULT, ROWSIZ, RQFILE, RQTAPE, SELEC1(30),
C      7 TTITLE(30), TOTPTS, VARSIZ, VECsiz,
C      8 CSEL(30), CSEL3(30), FETVC3(30), FETVEC(30), ZDUM,
C      9 ARRAY4(4000)
C
0004      REAL*8 ARRAY4
C
0005      REAL*4 CSET,CSET3,SELEC1,ANN1,ANN2
C
0006      INTEGER * 2 CSEL,CSEL3,FETVEC,FETVC3
C
0007      LOGICAL * 4 MAPFLG
C
0008      EQUIVALENCE (VECSIZ,NOFET3),(MTXSIZ,VARSIZ),(CELWTH,GRSIZE)
C
0009      REAL * 4 CELSUM(VECSIZ),FLOSUM(VECSIZ),AUXSUM(VECSIZ),
C      CELCOR(MTXSIZ),FLOCOR(MTXSIZ),
C      CELVAR(VECSIZ),FLOVAR(VECSIZ),AUXVAR(VECSIZ),
C      A1,A2,A3,A4,A5,A6,FDIST,DETERM,ALOG, SORT, FLOAT
C
C-----
C UNSUPERVISED MUV G.L.R. TEST
C MULTIPLE-UNIVARIATE TEST FOR EQUIVALENT MEAN VECTORS
C5 = FLOSIZ - 1
C8 = CELSIZ + FLOSIZ
120 J = 0
A1 = 1. + FDIST(1,C1+C5,ANN2)/(C1+C5)
DO 130 I=1,VECSIZ
J = J+1
AUXSUM(I) = CELSUM(I) + FLOSUM(I)
AUXVAR(I) = CELCOR(I) + FLOCOR(I) - AUXSUM(I)*AUXSUM(I)/C8
IF(AUXVAR(I) .GT. (CELVAR(I)+FLOVAR(I))*A1) RETURN
130 CONTINUE
C MULTIPLE-UNIVARIATE TEST FOR EQUIVALENT COVARIANCE MATRICES.
100 IF(ANN1 .LE. 1.E-25) GO TO 135
A1 = FLOAT(C1 + C5)
A2 = (1./C1 + 1./C5 - 1./A1)/3.
A3 = 3./(A2+A2)
A2 = (1. - A2 + 2./A3)/A3
A4 = FDIST(1,INT(A3*.01),ANN1)/A3
DO 110 I=1,VECSIZ
A3 = CELVAR(I)/C1
IF(A3 .LE. 0.) RETURN
A5 = FLOVAR(I)/C5
IF(A5 .LE. 0.) RETURN
A3 = A2*A1*ALOG(CELVAR(I)+FLOVAR(I)/A1)-C1*ALOG(A3)-C5*ALOG(A5)
IF(A3 .GE. 1.) RETURN
A5 = 1. - A3
IF(A3 .GT. A5*A4) RETURN
110 CONTINUE
135 DO 140 I=1,VECSIZ
FLOSUM(I) = AUXSUM(I)
140 FLOVAR(I) = AUXVAR(I)
DO 150 I=1,MTXSIZ
150 FLOCOR(I) = FLOCOR(I) + CELCOR(I)
FLOSIZ = C8
ANFLAG = 1
RETURN
END

```

JEC00010
JEC00020
JEC00030
JEC00040
JEC00050
JEC00060
JEC00070
JEC00080
JEC00090

JEC00260
JEC00270
JEC00280
JEC00290
JEC00300
JEC00310
JEC00320
JEC00330
JEC00340
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JEC00370
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JEC00670
JEC00680
JEC00690
JEC00700
JEC00710
JEC00720
JEC00730
JEC00740
JEC00750
JEC00760
JEC00770

A-42

LARS Program Abstract _____

MODULE IDENTIFICATION

Module Name: FDIST Function Name: NSIECHO

Purpose: Computes decision threshold of a F distribution

System/Language: CMS/FORTRAN

Author: _____ Date: _____

Latest Revisor: C. A. Pomalaza Date: 8/21/77

MODULE ABSTRACT

FDIST returns the decision threshold of the F distribution for a given significance level and a given number of degrees of freedom.

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West Lafayette, Indiana 47906

1. Module Usage

FDIST

CALL FDIST (NUMBER, DENOM, SIGLEV)

Arguments:

NUMBER I*4 degrees of freedom of the F distribution
DENOM

SIGLEV R*4 significance level

FDIST is a R*4 function.

2. Internal Description

FDIST looks at a table to get the value corresponding to the significance level. The significance level may be one of the following values:
.1, .05, .025, .01, .005, .001

3. Input Description

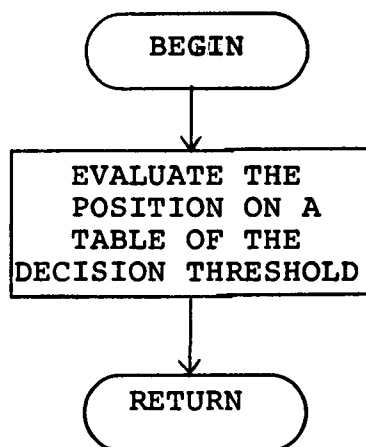
Not applicable

4. Output Description

Not applicable.

5. Supplemental Information

Not applicable.

6. Flowchart

LARS Program Abstract _____

MODULE IDENTIFICATION

Module Name: RDWRTE Function Name: NS1ECHO

Purpose: Writes intermediate tape

System/Language: CMS/FORTRAN

Author: _____ Date: _____

Latest Revisor: C.A. Pomalaza Date: 8/20/77

MODULE ABSTRACT

RDWRTE converts the disk files written by NSECHO into a intermediate tape file which is the input to NS1ECHO (the classification phase).

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1. Module Usage

RDWRTE

CALL RDWRTE (BDATA, RDATA, NROWPT, BUFFER, MEAN, COVAN)

All arguments but NROWPT are array bases computed by NSLINT.

BDATA	I*2 Array dimensioned as VECSIZ * ID(6) and used as in GADLIN (for raw data).
RDATA	R*4 Array dimensioned as NROWPT * VECSIZ and used as in GADLIN (holding calibrated data values from MIST).
NROWPT	I*4 Number of pixels per line + 6.
BUFFER	I*2 Array dimensioned JPTS * 4 where JPTS is the number of cell widths/line. It is used to read the flag buffer from disk.
MEAN	R*4 Array dimensioned as VECSIZ. Used to store cell means.
COVAN	R*4 Array dimensioned as VECSIZ * (VECSIZ + 1)/2. Used to store cell covariance matrices while processing.

2. Internal Description

RDWRTE performs the following functions:

1. The new ID record, the processing parameters and the field description card are written on the intermediate tape file using LARSYS System routine TOPWR.
2. The statistics file (mean and covariance) from each field are read from a disk file and written on tape using TOPWR.
3. The program starts a loop getting a line of data via LARSYS system routine GADLIN. It then checks the MAP option and if it is on, the appropriate flag line is read from disk, and the data read by GADLIN is altered to reflect field means if the pixels were annexed to fields. The means are read from the disk stat file. If the MAP option is off the data read by GADLIN is unaltered.
4. The line when finished is output to tape using TOPWR. The flag buffer line is also written using TOPWR.
5. The loop started in 3 continues until the area is finished.
6. An additional file (only the ID record) is written in the intermediate tape.

3. Input Description

Multispectral Data is read from the tape using GADLIN.

4. Output Description

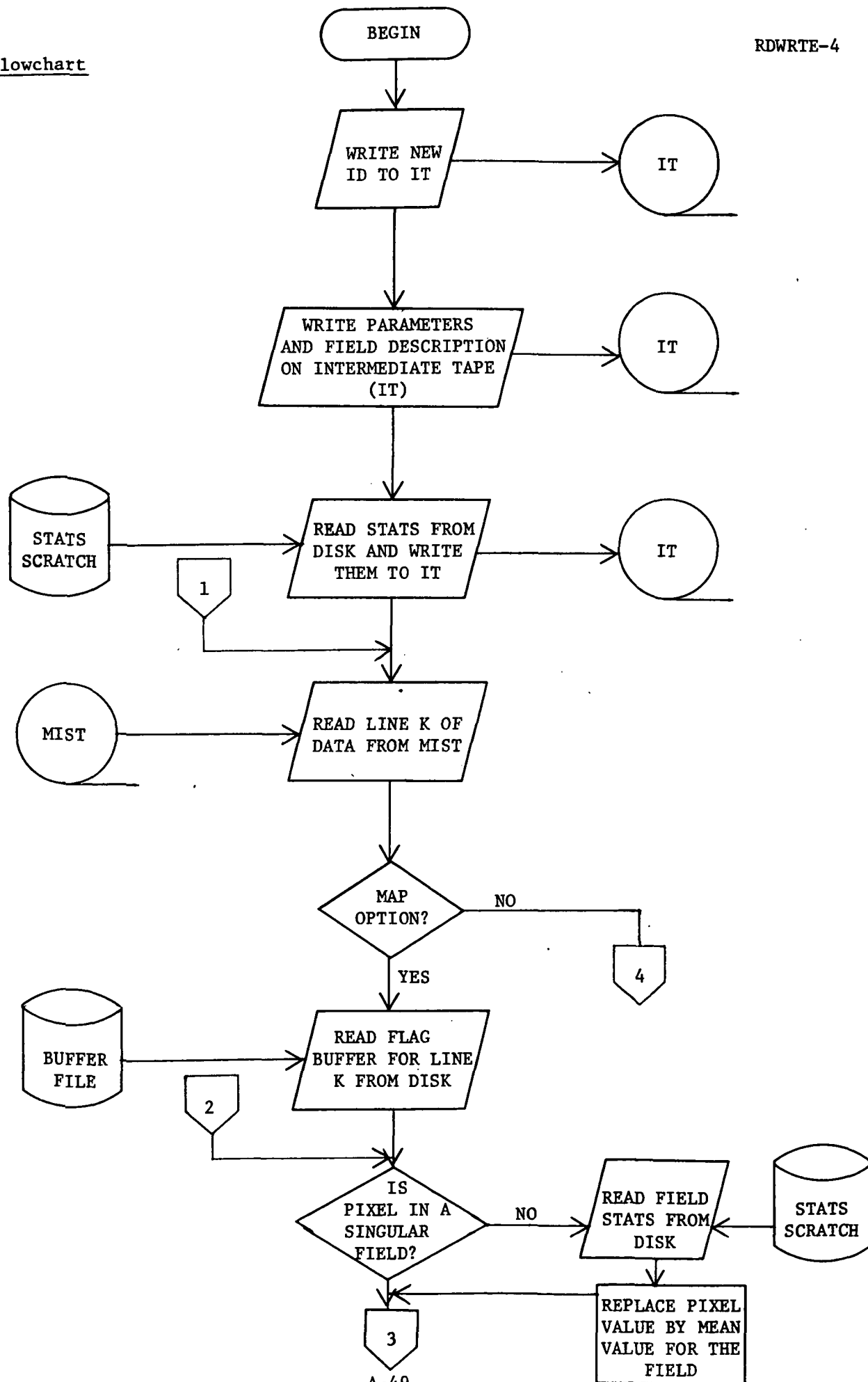
Intermediate tape for input to NS2ECHO (classification phase) is written.

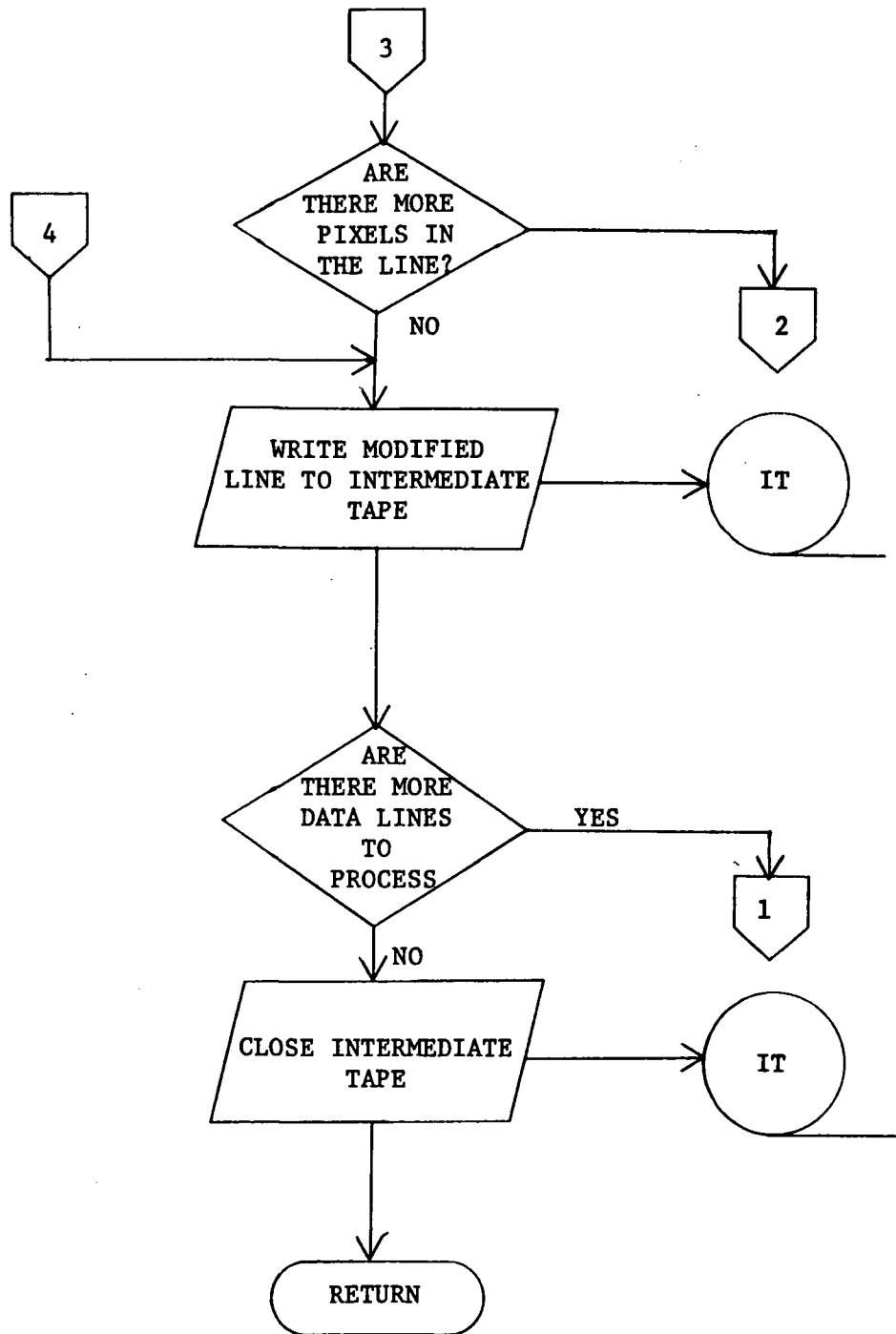
5. Supplemental Information

See intermediate tape file description.

6. Flowchart

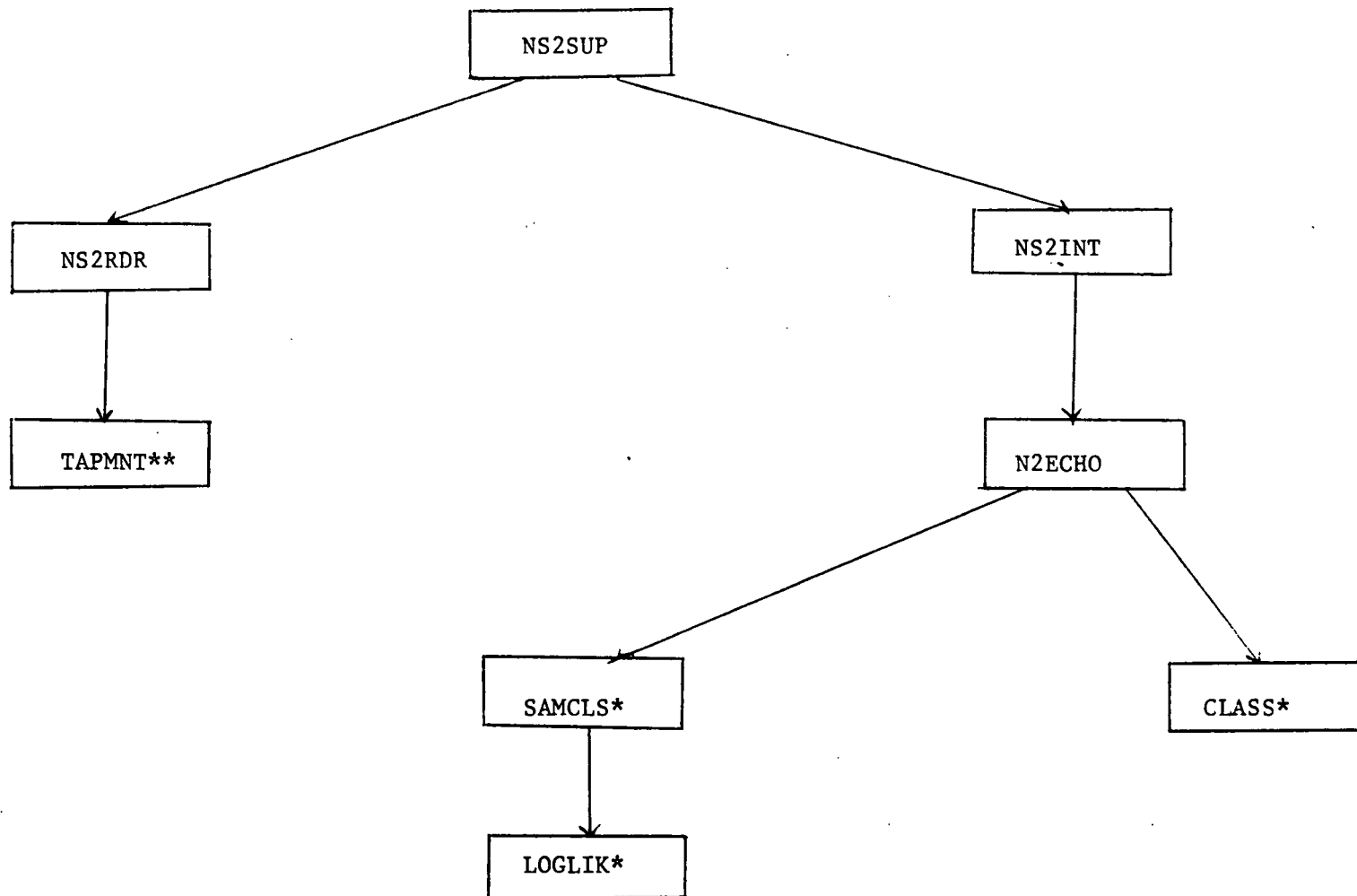
RDWRTE-4





[illegible]

MAIN SUBROUTINES TREE OF CLASSIFICATION PHASE
OF THE NONSUPERVISED ECHO PROCESSOR



*Delivered with Supervised ECHO documentation in the Final Technical Report on NASA
Contract NAS9-14970, May 31, 1977

**Delivered with Field Extraction Phase of the Nonsupervised ECHO processor

LARS Program Abstract_____

MODULE IDENTIFICATION

Module Name: NS2SUP Function Name: NS2ECHO

Purpose: Supervisor for NS2ECHO

System/Language: CMC/FORTRAN

Author: C. A. Pomalaza Date: 8/21/77

Latest Revisor: _____ Date: _____

MODULE ABSTRACT

NS2SUP receives control from LARSMN. This supervisor performs no computation, but instead makes call to the subroutines which really make up the processor.

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1. Module Usage

NS2SUP

There are no parameters to NS2SUP. It is called by LARSMN when NS2ECHO is requested. Control returns to LARSMN when the function is completed.

2. Internal Description

NS2SUP first calls the subroutine NS2RDR to read in all the control cards, and then calls NS2INT to read the statistics data and compute the array bases. NS2INT calls NS2ECHO which performs the classification.

3. Input Description

Not applicable.

4. Output Description

Two messages are produced and written to unit TYPEWR (the console).

UNSUPERVISED ECHO FUNCTION (PHASE 2) REQUESTED

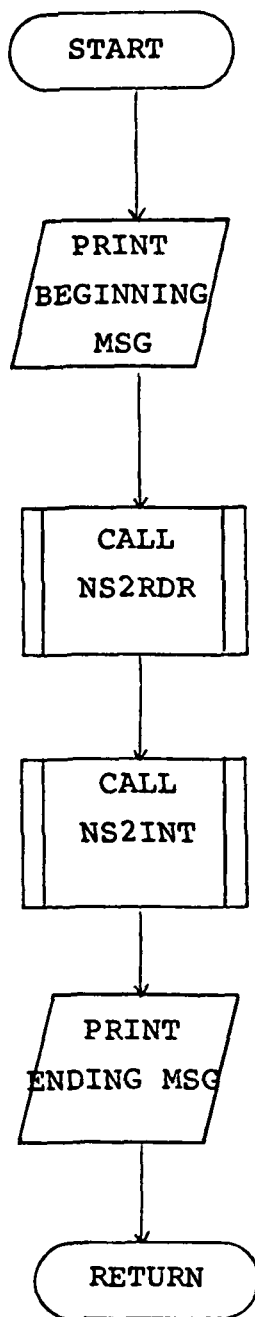
Signifies beginning of the function.

UNSUPERVISED ECHO FUNCTION (PHASE 2) COMPLETED

Signifies end of function.

5. Supplemental Information

Not applicable.

6 - Flowchart

FILE NS2SUP

```

C      NS2SUP  LARS XXXX
C      WRITTEN BY C.A. POMALAZA
C      *****
0001      SUBROUTINE NS2SUP
C
0002      IMPLICIT INTEGER*4 (A-Z)
0003      COMMON /GLCCOM/ BLANK, CARD(20), CHKOUT, COPFIL, CLASSR, CLASSX,
1      CLUSTX, CONPUT, CPYOUT, CRDRDR, CROSEQ, DATAPE,
2      DUPLTP, DUPRUN, ERRMSG, FBPNT,
3      FILESV, FLDBND, HDATA, HEAD(88), ID(200), IMAGEX,
4      IMARK, KEYBD, MAPTAP, MAXCHA, MAXCLS,
5      PAGESZ, PNCH, POINT, PRESUX, PRNTR, READIN,
6      RESTR, RUNFIL, RUNTAB(10,3),
7      SDATA, SEPARX, SEPTPX, SPARE(10), TEMPAS(30),
8      TPSTAT(6), TTFLDX, TYPEWR,
9      TOP, ARRAY(12500)
0004      REAL * 8 ARRAY
0005      REAL * 4 FRQCAL(5,30)
0006      INTEGER * 4 COMENT(16), DATE(5), HED1(16), HED2(16), TIME(5)
0007      INTEGER * 2 BLANK2, NNID(400)
0008      LOGICAL * 4 CHKOUT
0009      LOGICAL * 1 BLANK1
0010      EQUIVALENCE (DATSAV, ID(1)), (CURRIN, ID(3)), (FRQCAL(1), ID(51)),
1      (HED1(1), HEAD(3)), (DATE(1), HEAD(26)), (HED2(1), HEAD(39)),
2      (TIME(1), HEAD(58)), (COMENT(1), HEAD(72)),
3      (MAPSAV, TPSTAT(1)),
4      (SEPSCR, TPSTAT(2)), (DUPIN, TPSTAT(3)), (DASTAT, TPSTAT(4)),
5      (COPSER, TPSTAT(5)), (TRAOUT, TPSTAT(6)),
6      (BLANK, BLANK2, CLANK1)
C
0011      COMMON /NS2COM/ CSET(3,30), MTXSIZ, NOFET3, NOPOOL, OFILE, OTAPE,
1      POLNAM(2,60), RESULT, RQFILE, RQTAPE, STKPTR, VECsiz, WRKSIZ,
2      CSEL(30), POLPTR(2,60), POLSTK(60), POLNM1(60)
C
0012      REAL*4 CSET
0013      INTEGER*4 POLNAM
0014      INTEGER*2 CSEL, POLPTR, POLSTK
0015      LOGICAL*1 POLNM1
C
0016      LOGICAL*1 POLNM1
C
C      *****
C      CALL READER AFTER INITIAL MESSAGE PRODUCED
C      *****
0017      WRITE(TYPEWR, 100)
0018      100 FORMAT(' 1000 UNSUPERVISED ECHO FUNCTION (PHASE 2)',
1      * ' REQUESTED (NS2SUP)' )
C
0019      CALL NS2RDR
C
C      *****
C      CALL PROCESSING ROUTINE THEN PRINT TERMINATION MESSAGE
C      *****
0020      CALL NS2INT
C
0021      WRITE(TYPEWR, 9000)
0022      9000 FORMAT(' 10000 UNSUPERVISED ECHO FUNCTION (PHASE2) COMPLETED ',
1      * '(NS2SUP)' )
C
0023      RETURN
0024      END

```

```

*NS100010
*NS100020
*NS100030
*NS100040
*NS100050
*NS100060
*NS100070
*NS100080
*NS100090
*NS100100
*NS100110
*NS100120
*NS100130
*NS100140
*NS100150
*NS100160
*NS100170
*NS100180
*NS100190
*NS100200
*NS100210
*NS100220
*NS100230
*NS100240
*NS100250
*NS100260
*NS100270
*NS100280
*NS100290
*NS100300
*NS100310
*NS100320
*NS100330
*NS100400
*NS100410
*NS100420
*NS100430
*NS100440
*NS100450
*NS100460
*NS100470
*NS100480
*NS100490
*NS100500
*NS100510
*NS100520
*NS100530
*NS100540
*NS100550
*NS100560
*NS100570
*NS100580
*NS100590
*NS100600
*NS100610
*NS100620
*NS100630
*NS100640
*NS100650
*NS100660

```

LARS Program Abstract_____

MODULE IDENTIFICATION

Module Name: NS2RDR Function Name: NS2ECHO

Purpose: Read functions control card

System/Language: CMS/FORTRAN

Author: C. A. Pomalaza Date: 8/21/77

Latest Revisor: _____ Date: _____

MODULE ABSTRACT

NS2RDR reads and interprets all function control cards for NS2ECHO. Also a results tape is readied.

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1. Module Usage

NS2RDR

CALL NS2RDR

This section lists the actions taken when the following control cards are read:

RESULTS---TAPE	- The variable RQTAPE is set to the given tape number.
RESULTS---FILE	- The variable RQFILE is set to the given file number.
RESULTS---INITIALIZE	- The local flag INITFG is set to TRUE. This flag is used by NS2RDR to call MMTAPE to initialize a tape.
INTERMEDIATE---TAPE	- The variable OTAPE is set to the given tape number.
INTERMEDIATE---FILE	- The variable OFILE is set to the given file number.
CLASSES	- LARSYS System Subroutine POLSCN is called to set up the arrays POLNAM, POLPTR, POLSTK, and POLNMI and compute NOPPOL and STKPTR based on the interpretation of the classes card.

2. Internal Description

NS2RDR uses standard card reader logic in using LARSYS system routines CTLWRD, CTLPRM, and IVAL in reading and interpreting the control cards. After initializing flags and arrays which will convey control card information, NS2RDR functions in a loop of reading and interpreting control cards until DATA or END card is read indicating the end of function control card. After control cards have been read some checks are made on the data. If the classes are pooled, the pooling is checked for validity. Then the results tape is mounted and, if requested, initialized.

3. Input Description

Function control cards for NS2ECHO are read via LARSYS System routine CTLWRD.

4. Output Description

Control card error messages are written to both the printer (PRNTR) and the console (TYPEWR). A brief list of these follows:

ERROR IN CLASSES CARD. CORRECT ALL CLASSES CARDS AND START OVER.

ERROR IN RESULTS CARD (TAPE OR FILE PARAMETER) - TYPE IN CORRECT CARD.

Syntax error in the TAPE or FILE specification for the intermediate or results tape. Standard corrective action.

A POOL HAS NOT BEEN DEFINED. CORRECT CLASSES CARD. POOL NUMBER IS

The pool number is written on the next line. The function terminates. Pool numbers must be consecutive and start at 1.

BOTH FILE AND INITIALIZATION OPTION REQUESTED. FILE REQUESTED IGNORED. FUNCTION CONTINUES.

The results tape is initialized. Only file 1 can be initialized.

UNEXPECTED END OF FILE ON INPUT DATA

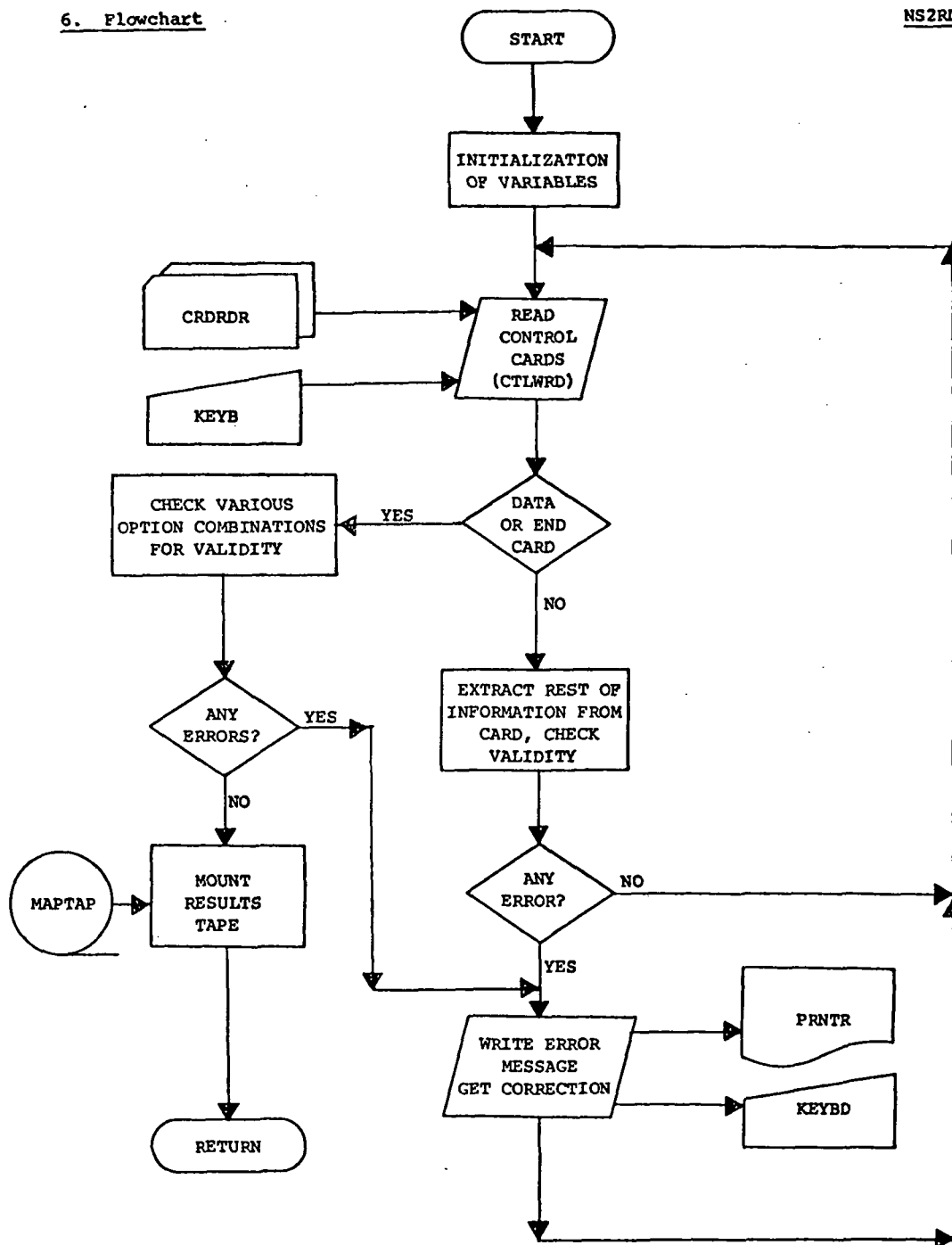
The end of the input deck was reached while reading cards for the function. This normally means that the END card was omitted.

5. Supplemental Information

See LARSYS System Manual for card reader requirements.

6. Flowchart

NS2RDR



[illegible]

PAGE 0002

[illegible]

FORTRAN IV C LEVEL 20.7 NS2RDRL DATE = 77224 11052000 PAGE 0003
FILE NS2RDR

```

C      INITIALIZE SPECIFICATION
C      130 INITFG = .TRUE.
C      0073 GO TO 300
C      0074
C      *****
C      INTERMEDIATE TAPE SPECIFICATIONS
C      *****
C      400 IFICOL .EQ. 723 GO TO 100
C      CALL CILDRN(CARD,COL, RESCOD,2,CUCF, C111)
C      0075 IFVICSZ .EQ. 01 GO TO 415
C      0076 GO TO 400
C      *****
C      TAPE SPECIFICATIONS
C      410 VICSZ = 1
C      CALL IVALICARD, COL, OTAPL, VICSZ, 6415)
C      0078 IFVICSZ .EQ. 01 GO TO 415
C      0079 GO TO 400
C      415 TRNNUM = 148
C      0080 GO TO 110
C      0081
C      *****
C      FILE SPECIFICATION
C      420 VICSZ = 1
C      CALL IVALICARD, COL, OTAPL, VICSZ, 6415)
C      0084 IFVICSZ .EQ. 01 GO TO 415
C      0085 GO TO 400
C      0086
C      *****
C      CHECK FOR PRESENCE AND VALIDITY OF ALL INFORMATION IN CARD
C      *****
C      500 IF(STATUS .EQ. 1) CALL RTMAIN
C      0088 IF (NDPOOL .LE. 0) GO TO 540
C      0089 GO TO 525
C      520 IF (POLPTR(1,1) .EQ. 0) GO TO 540
C      0090
C      525 CONTINUE
C      0091 GO TO 540
C      0092 CALL ERPRINT(344, 'GO')
C      0093 WRITE (PRINT, 9530) 1
C      0094 WRITE (TYPEWR, 9530) 1
C      0095
C      950 FORMAT ('14')
C      0096 CALL RTMAIN
C      0097
C      *****
C      CHECK RESULTS, CARD ENTRIES, ANY FILE ENTRY WITH RESULTS
C      *****
C      540 IF (.NOT. INITFG) GO TO 545
C      0099 IF (RUFIL .EQ. 0) GO TO 542
C      0100 WRITE (TYPEWR, 9440)
C      0101 WRITE (PRINT, 9540)
C      0102
C      9540 IF (FILE REQUESTED IGNORED, FUNCTION CONTINUES INS2RDR)
C      0103
C      542 RUFIL = 1
C      0104 CALL RMTAPE(ROTAPL, 0, 1)
C      0105 GO TO 500
C      545 CALL RMTAPE(ROTAPL, RUFIL, 1)
C      0106 IF (RUFIL .EQ. 0) GO TO 900
C      0107 ASSIN 540 TO GOVEC
C      0108
C      900 ERROR = 1
C      0109 RUFIL = 1 ABS(RUFIL)
C      0110 GO TO 100
C      0111
C      400
C      0112
C      0113
C      0114

```

LARS Program Abstract

MODULE IDENTIFICATION

Module Name: NS2INT

Function Name: NS2ECHO

Purpose: Initialization including array bases computation

System/Language: CMS/FORTRAN

Author: C. A. Pomalaza

Date: 8/20/77

Latest Revisor:

Date:

MODULE ABSTRACT

NS2INT carries out required initialization of the rest of the variables used by NS2ECHO and finishes reading the statistics to be used by the processor. Also the intermediate tape is readied and the subroutine that performs the classification is called.

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1. Module Usage

NS2INT

CALL NS2INT

The program NS2INT is called with no arguments. Any variables to be used or changed are contained in common blocks GLOCOM and NS2COM.

2. Internal Description

NS2INT performs the following functions:

1. Uses LARSYS system routine STAT to read the statistics from cards.
2. Computes space needed for arrays to hold original statistics.
3. Use LARSYS system routine CLSCHK to check class validity.
4. The intermediate tape is mounted and positioned via TAPMNT.
5. With the information read from the intermediate tape it computes array bases for reduced statistics.
6. Reduce statistics using LARSYS system routine REDSAV.
7. The first three record types are written on the results file. The first record has a 0 in the sixth full word to indicate the absence of weights in the file.
8. Some processor information is printed out.
9. With the information from the Intermediate tape the array bases for calling N2ECHO are computed.
10. N2ECHO is called for performing the classification when finished it writes the needed tape marks and the check record.

3. Input Description

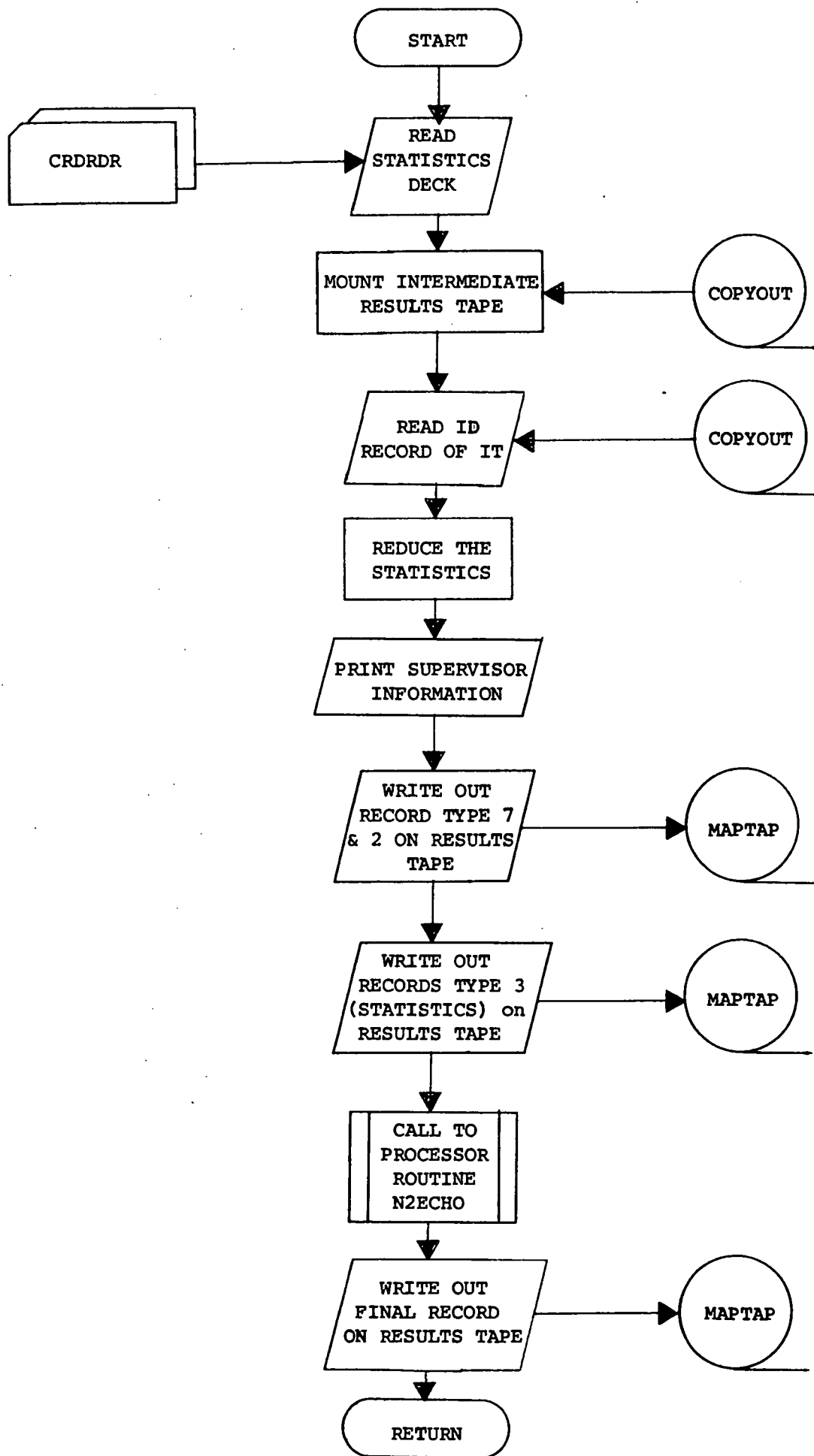
Statistics are read via a call to LARSYS system routine STATS. Information from the ID record of the intermediate tape is obtained via a call to LARSYS system routine TOPRD.

4. Output Description

Information concerning serial number, classes, field and channels is written on the printer (PRNTR). The results tape is written using unformatted FORTRAN write statement.

5. Supplemental Information

Not applicable.



[illegible][illegible]

LARS Program Abstract_____

MODULE IDENTIFICATION

Module Name: N2ECHO Function Name: NS2ECHO

Purpose: Performs classification for NS2ECHO and writes results tape.

System/Language: CMS/FORTRAN

Author: _____ Date: _____

Latest Revisor: C. A. Pomalaza Date: 8/21/77

MODULE ABSTRACT

N2ECHO is called by NS2INT to perform the field classification on those fields and singular points identified by NS2ECHO (field extraction phase) and wirtten on an intermediate tape. N2ECHO writes a standard results file to tape.

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1. Module UsageN2ECHO

CALL N2ECHO (COVMTX, AVEMTX, SCAPRD, DETCOV, CONST, WORK, MEAN, COV, SAVAVE, HOLD, CHISQR, RDATA, DATBS3)

Input Arguments

COVMTX	R*1 Array of covariances obtained in NS2INT. Dimensioned NTXSIZ*NOPOOL where NOPOOL is the number of classes contained in the reduced STAT DECK and MTXSIZ = VECSIZ * (VECSIZ + 1)/2 where VECSIZ is the number of channels requested for classification (in Phase 1).
AVEMTX	R*4 Array of mean vectors for each class. Dimensioned VECSIZ*NOPOOL.
SCAPRD	R*4 Array of the scalar product of the mean and covariance matrices for each class dimensioned NOPOOL.
DETCOV	R*4 Array of the determinant of the covariance matrix for each class dimensioned NOPOOL.
CONST	R*4 Array of the constant terms used in the computation of the likelihood values dimensioned NOPOOL.
WORK	R*4 Working array for a subroutine cell dimensioned VECSIZ * (VECSIZ + 2).
MEAN	R*4 Array used as a holding buffer for a subroutine cell. Dimensioned VECSIZ.
COV	R*4 Array used as a holding buffer for a subroutine cell. Dimensioned MTXSIZ.
SAVAVE	R*4 Array used for saving the mean from each class of the STAT deck. Dimensioned VECSIZ*NOPOOL.
HOLD	R*4 Array used as a holding buffer for subroutine call. Dimensioned 7*VECSIZ.
CHISQR	R*4 Array used in a subroutine call. Dimensioned NOPOOL.
RDATA	R*4 Area to be used for dynamic allocation. This must be the first unused element of ARRAY in the calling program.

DATBS3 I*4 Number of bytes on ARRAY which one in use
 (i.e. the number of bytes in ARRAY which precede
 RDATA).

Output Arguments

Not applicable.

2. Internal Description

- 1 - N2ECHO calls to LARSYS system routine SMMULT and SAMINV to invert the covariance matrices in COVMTX and produce the determinant in DETCOV needed for classification.
- 2 - A loop begins where each field statistics matrix (record type 4) on the intermediate tape is classified by LARSYS system routine SAMCLS. The absolute field number (see tape record description) indexes the class.
- 3 - The records that associates each pixel with its appropriate field are read. The processor classifies each pixel in an homogeneous field by looking up its absolute field number. Singular pixels are classified by a call to CLASS. The results are written line by line following the standard LARSYS format (see RESULTS FILE description in LARSYS System Manual).
- 4 - The RESULTS FILE is closed and control returns to NS2INT.

3. Input Description

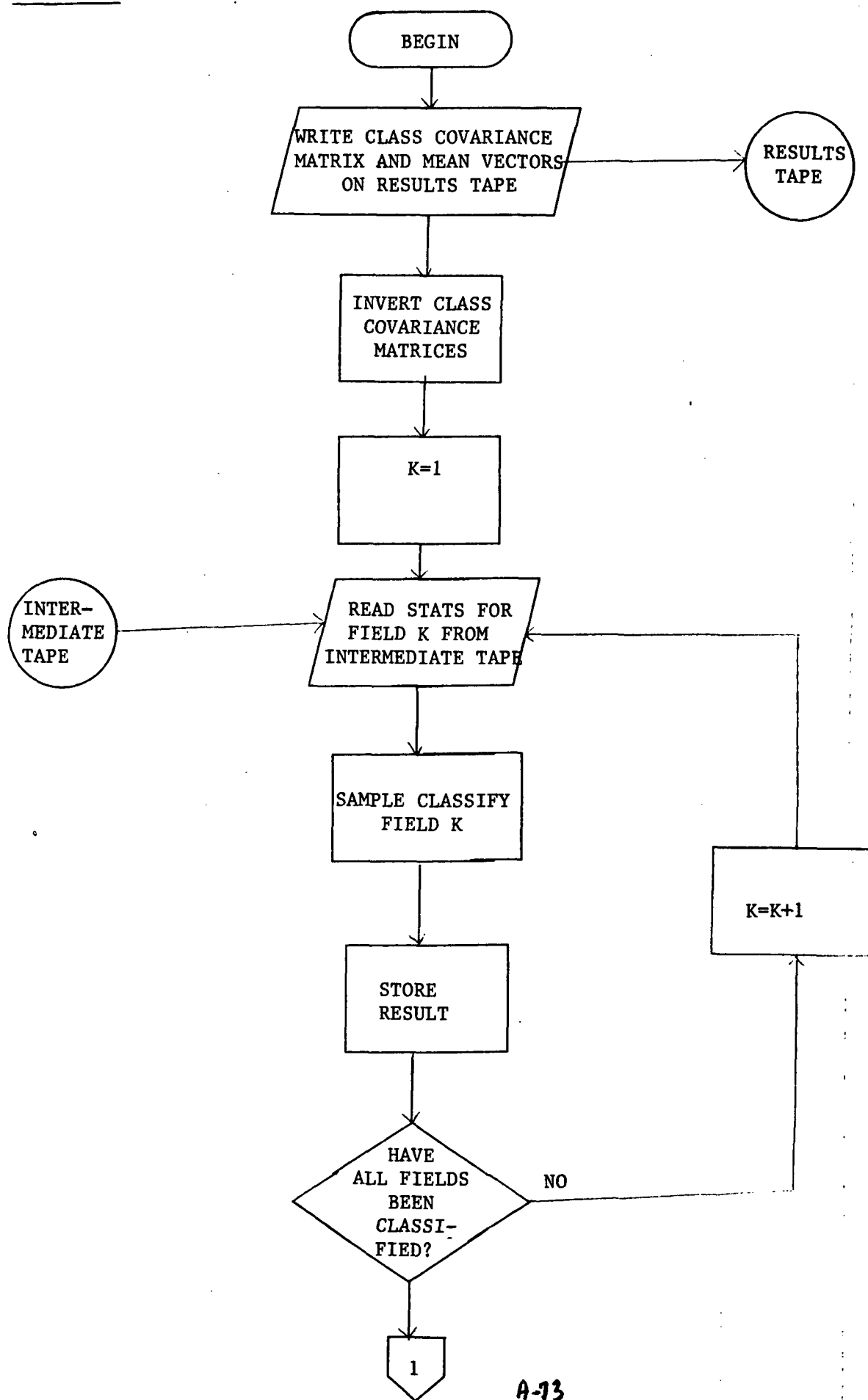
N2ECHO reads field statistics information and pixel data from the intermediate tape via call to LARSYS system routines TOPRD, and GADLIN.

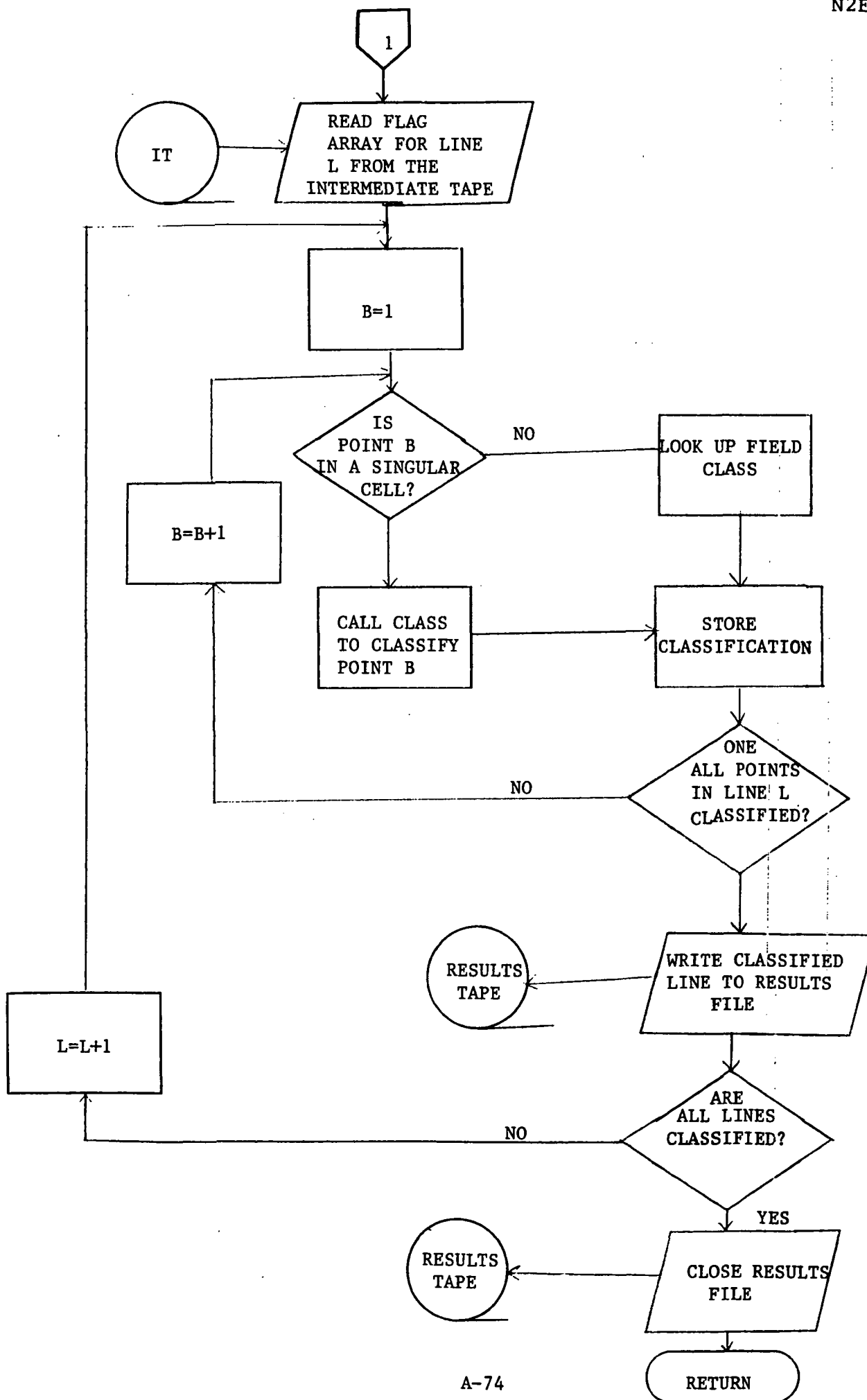
4. Output Description

A standard results file is output to tape. See LARSYS System Manual for format description.

5. Supplemental Information

See non-supervised ECHO intermediate tape description. Also Results Tape description on LARSYS System Manual.

6. Flowchart



A-75

```

C
C N2ECHO LARS XXXX
C WRITTEN BY C. A. POMALAZA
C *****
0001 SUBROUTINE N2ECHO(COVMTX,AVEMTX,SCAPRD,DETCOV,CONST,WORK,
C & MEAN,COV,SAVAVE,HOLD,CHISQR,RODATA,DATBS3)
0002 IMPLICIT INTEGER*4 (A-Z)
0003 COMMON /GLCCOM/ BLANK, CARD(20), CHKOUT, COPFIL, CLASSR, CLASSX,
1 CLUSTX, CONPUT, CPYOUT, CRDRDR, CRDSEQ, DATAPE,
2 DUPLTP, DUPRUN, ERRMSG, FRPNT,
3 FILESV, FLCBND, HDATA, HEAD(88), ID(200), IMAGEX,
4 IMARK, KEYBD, MAPTAP, MAXCHA, MAXCLS,
5 PAGESZ, PNCN, POINT, PRESUX, PRNTR, READIN,
6 RESTAT, RUNFIL, RUNTAB(10,3),
7 SDATA, SEPARX, SEPTPX, SPARE(10), TEMPAS(30),
8 TPSTAT(6), TTFLOX, TYPEWR,
9 TOP, XARRAY(12500)
0004 REAL * 8 ARRAY
0005 REAL * 4 FROCAL(5,30)
0006 INTEGER * 4 COMMENT(16), DATE(5), HED1(16), HED2(16), TIME(5)
0007 LOGICAL * 4 CHKOUT
0008 INTEGER * 2 BLANK2
0009 LOGICAL * 1 BLANK1
0010 EQUIVALENCE (DATSAV,ID(1)), (CURRUN,ID(3)), (FROCAL(1),ID(51)),
1 (HED1(1),HEAD(81)), (DATE(1),HEAD(26)), (HED2(1),HEAD(39)),
2 (TIME(1),HEAD(58)), (COMMENT(1),HEAD(72)),
3 (MAPSAV,TPSTAT(1)),
4 (SEPSER,TPSTAT(2)), (DUPIN,TPSTAT(3)), (DASTAT,TPSTAT(4)),
5 (CUPSER,TPSTAT(5)), (TRADUT,TPSTAT(6)),
6 (BLANK,BLANK2,BLANK1)
0011 COMMON /NS2COM/ CSET(3,30), MTXSIZ, NOFET3, NOPOOL, OFILE, OTAPE,
1 POLNAM(2,60), RESULT, ROFILE, ROTAPE, STKPTR, VECsiz, WRKSIZ,
2 CSEL(30), POLPTR(2,60), POLSTK(60), POLNMI(60)
0012 REAL*4 CSET
0013 INTEGER*4 POLNAM
0014 INTEGER*2 CSEL, POLPTR, POLSTK
0015 LOGICAL*1 POLNMI
0016 LOGICAL*1 POLNMI
C
C DEFINITION OF LOCAL VARIABLES
0017 REAL * 4 SPAR(10),AR(125)
0018 REAL* 4 CSET(3,30)
0019 INTEGER* 4 AR1(125)
0020 INTEGER* 2 A,B,AR2(250)
0021 INTEGER*2 CSEL(30), BLOCK(4)
0022 LOGICAL* 1 LR(8), RS(2)
0023 EQUIVALENCE (AR(1),AR1(1)), AR2(1))
0024 EQUIVALENCE (LR(1), LR1(1))
C
0025 REAL*4 MEAN(VECSIZ),COV(MTXSIZ)
0026 INTEGER*4 REJBUE(20),IR(2)
0027 INTEGER*2 DIMFLG,NEGONE,IN,FETVEC(30),
& FETVC(30)
0028 EQUIVALENCE (IN,RS(1))
0029 INTEGER*4 INFO(17),PREFIX(2)
0030 INTEGER*2 BUFFER(2500)
0031 INTEGER * 2 SUBUF(22500)
0032 REAL*4 HDATA(1)
0033 INTEGER*2 TAPBUF(1000)
0034 REAL*4 COVMTX(MTXSIZ,NOPOL),AVEMTX(VECSIZ,NOPOL),WORK(WRKSIZ),
& SCAPRD(NOPOL),CHISQR(NOPOL),DETCOV(NOPOL),CONST(NOPOL),
& HOLD(7,VECSIZ), SAVAVE(VECSIZ,NOPOL), THRTAB(234), PCREJ
C

```

```

*GCC00010
GCC000120
GCC000130
GCC000140
GCC000150
GCC000160
GCC000170
GCC000180
GCC000190
GCC000200
GCC000210
GCC000220
GCC000230
GCC000240
GCC000250
GCC000260
GCC000270
GCC000280
GCC000290
GCC000300
GCC000310
GCC000320
GCC000330
GCC000340
GCC000350
*GCC000450
GCC000460
GCC000470
GCC000480
GCC000490
GCC000500
GCC000510
GCC000520
GCC000530
GCC000540
GCC000550
GCC000560
GCC000570
GCC000580
GCC000590
GCC000600
GCC000610
GCC000620
GCC000630
GCC000640
GCC000650
GCC000660
GCC000670
GCC000680
GCC000690
GCC000700

```

```

0035      WRITE COVARIANCE MATRIX AND MEAN VECTOR ON RESULTS TAPE
0036
0037      REWIND SDATA
0038      PREFIX(1)=4
0039      PREFIX(2)=1
      WRITE(RESULT)PREFIX,COVMTX,AVEMTX
      NEGOCIE=-1
      INVERT THE LEADING STAT DECKS ON INTERMEDIATE TAPE
      DO 510 I=1,NOPPOOL
      CALL SAMT4V(COVMTX(I,1),NOFET3,DETCOV(I),WORK)
      IF (DETCOV(I).LE.0.) GO TO 520
      DO 788 N = 1, VECsiz
      788 SAVAVE(N,I) = AVEMTX(N,I)
      CALL SMMULT(COVMTX(I,1),AVEMTX(I,1),SCAPRD(I),WORK,NOFET3)
      510 CONST(I)=-5*ALOG(16.2831853**NOFET3)*DETCOV(I)
      THE CODE BELOW BUILDS A TABLE OF THRESHOLD VALUES
      CORRESPONDING TO FRACTIONS REJECTED FROM 0.1 PERCENT
      TO 99 PERCENT. PCREJ IS 1.0 -PERCENT REJECTION/100.
      PCREJ = 1.0
      DO 511 I=1,234
      PCREJ = PCREJ - .001
      IF(I.GT.150) PCREJ = PCREJ - 0.009
      CALL TRESJC(PCREJ,THRTAB(I),VECSIZ,ERROR)
      IF(THRTAB(I).NE.0) CALL ERPRNT(377,'STOP')
      511 CONTINUE
      GO TO 515
      520 CALL ERPRNT(160,'GO')
      WRITE(PRINTER,9222)
      9222 FORMAT('SINGULAR CLASS')
      STOP
      CLASSIFY EACH FIELD
      535 CALL TCRPD(12,68,ERROR,INFO,2,WLF)
      CALL TOPRD(12,16,ERROR,SPAR,2,WLF)
      CCLWTH = SPAR(1)
      LCOUNT=8*(VECSIZ+MTXSIZ)*4
      L1=10(4)
      DO 20 J=1,L1
      CALL TCRPD(12,LCOUNT,ERROR,AR(1),2,WLF)
      A=AR2(1)
      B=AR2(2)
      D=AR1(2)
      DO 11 K2=1,VECSIZ
      11 MEAN(K2)=A.(2+K2)
      DO 12 K2=1,MTXSIZ
      12 COM(K2)=AR1(2+VECSIZ+K2)
      CALL SAMCLSCCLASSI,CHISQR,A1,COVMTX,AVEMTX,SCAPRD,CONST,COV,MEAN,
      & D,VECSIZ,MTXSIZ,NOPPOOL,.FALSE.)
      SORTU(J)=CLASSI
      20 CONTINUE
      WRITE RECORD 5 ON RESULTS FILE
      PREFIX(1)=5

```

```

0077      PREFIX(2)=1                      GCC001470
0078      PTQ=(INFO(3)-INFO(7)+1)/INFO(9)    GCC001480
0079      LINQ=(INFO(5)-INFO(4)+1)/INFO(6)    GCC001490
0080      WRITE(RESULT)PREFIX,PTQ,LINQ,INFO,1D,((CSET(I,J),I=1,3),J=1,30) GCC001500
*****
      CHECK FOR CORE OVERFLOW              *GCC001510
*****                                     *GCC001520
      NSR = PTQ + 6                         *GCC001530
      NSD = ID(6)                           *GCC001540
      RBASE = NSR*VECSIZ + 1                *GCC001550
      LB = (NSD*VECSIZ + 3)/4 + RBASE        *GCC001560
      IF (MOD(LB,2) .EQ. 1) LB = LB + 1      *GCC001570
      BASE = DAT(5)*8 + LB*4                *GCC001580
      CORE = TOP - BASE                     *GCC001590
      IF(CORE .GT. 0) GO TO 650              *GCC001600
      CORE = -CORE                           *GCC001610
      CALL ERPRNT(376, 'GO')                 *GCC001620
      WRITE(1TYPEWR, 9600) CORE              *GCC001630
      WRITE(1PTR, 9600) CORE                 *GCC001640
      FORMAT(10X, 16, 'BYTES')              *GCC001650
      CALL RTMAIN                             *GCC001660
*****                                     *GCC001670
      ENTER MAIN LOOP OF CLASSIFICATION      *GCC001680
*****                                     *GCC001690
0095      DO 600 I = 1,30                   *GCC001700
0096      CSET(I,1) = 0                      *GCC001710
0097      DO 600 J = 1,3                     *GCC001720
0098      CSET3(J,I) = -50000.                *GCC001730
0099      K = 0                               *GCC001740
0100      DO 610 I = 1,30                     *GCC001750
0101      IF(CSEL(I) .EQ. 0) GO TO 610        *GCC001760
0102      K = K + 1                           *GCC001770
0103      CSEL3(K) = CSEL(I)                  *GCC001780
0104      DO 620 J = 1,3                      *GCC001790
0105      CSET3(J,K) = FRQCAL(J+2, I)          *GCC001800
0106      CONTINUE                             *GCC001810
0107      CONTINUE                             *GCC001820
*****                                     *GCC001830
      BLOCK(2) = 1                           *GCC001840
      BLOCK(3) = PTQ                          *GCC001850
      BLOCK(4) = 1                            *GCC001860
      PTS = PTQ                               *GCC001870
      PTS6 = PTS + 6                          *GCC001880
      JPTS=PTS/CELWTH                         *GCC001890
      BUF=JPTS                                *GCC001900
      COUNT=VECSIZ*(PTS6)+4                  *GCC001910
      DO 690 KM=1,LINQ                        *GCC001920
      BLOCK(1) = KM                           *GCC001930
      CALL GADLIN(BLOCK(1), CSEL3, CSET3, ID, 12, VECSIZ, NSR, *GCC001940
      RDATA(BASE), RDATA(1), RCLL, ERROR)    *GCC001950
      RCPTR=1                                 *GCC001960
      BUFPTR=1                                *GCC001970
      OUTPTR=3                                *GCC001980
      TAPBUF(1) = 0                           *GCC001990
      TAPBUF(2)=KM*INFO(4)-1                  *GCC002000
      DO 700 K9=1,VECSIZ                      *GCC002010
      DO 700 K10=1,6                           *GCC002020
      FCLO(K10+1, K9) = RDATA( PTS +(K9-1)*PTS6 + K10) *GCC002030
      DO 700 KCO=1,JPTS                         *GCC002040
      IF (BUFFER(BUFPTR).EQ.NEGONE) GO TO 720 *GCC002050
      DO 710 K2=1,CELWTH                       *GCC002060
      BUFRUM=BUFFER(BUFPTR+1)                  *GCC002070
      TAPBUF(OUTPTR)=SOBUF(BUFRUM)            *GCC002080
      710 OUTPTR=OUTPTR+1                       *GCC002090
      ROPTR=ROPTR+CELWTH                       *GCC002100
      BUFPTR=BUFPTR+4                          *GCC002110
      703                                     *GCC002120
      690                                     *GCC002130
      650                                     *GCC002140
      CONTINUE                                *GCC002150
      CONTINUE                                *GCC002160
      CONTINUE                                *GCC002170
      CONTINUE                                *GCC002180
      CONTINUE                                *GCC002190
      CONTINUE                                *GCC002200
      CONTINUE                                *GCC002210
      CONTINUE                                *GCC002220

```


FILE N2ECHO

```

0136      GO TO 700
0137 720 DO 740 K7=1,CELWTH
0138      DO 730 K3=1,VECSIZ
0139 730 HCOLD(1,K3)=RDATA(RDPTR + (K3-1)*PTS6 )
0140 777 CALL CLASS(HOLD,VECSIZ,1,NPOOL,SAVAVE,COVMTX,IR(1),IR(2),CONST,
      & THRTAB)
0141      RS(2)=IR(2)
0142      RS(1)=IR(6)
0143      TAPBUF(OUTPTR)=IN
0144      OUTPTR=OUTPTR+1
0145 740 RDPTR=RDPTR+1
0146      BUFPTR=BUFPTR+4
0147 700 CONTINUE
0148      PREFIX(1)=6
0149      PREFIX(2)=1
0150      PT=PTS+2
0151      WRITE(11)PREFIX,(TAPBUF(J),J=1,PT)
0152 692 CONTINUE
0153      TAPBUF(1)=0
0154      TAPBUF(2)=0
0155      PREFIX(1)=7
0156      WRITE(RESULT)PREFIX,(TAPBUF(J),J=1,PT)
0157      PTQ = 6
0158      PREFIX(1) = 8
0159      LINO = 0
0160      WRITE(RESULT)PREFIX,PTQ,LINO,INFO,IO,CSET
0161      RETURN
0162      END

```

```

GCC02230
GCC02240
GCC02250
GCC02260
GCC02270
GCC02280
GCC02290
GCC02300
GCC02310
GCC02320
GCC02330
GCC02340
GCC02350
GCC02360
GCC02370
GCC02380
GCC02390
GCC02400
GCC02410
GCC02420
GCC02430
GCC02440
GCC02450
GCC02460
GCC02470
GCC02480
GCC02490
GCC02500

```

APPENDIX B

Ancillary Data Files Required
by the
Nonsupervised ECHO Processor

Nonsupervised Field Extraction Processor Disk Files

The NSLECHO processor writes two disk files during execution. The first file BUFFER FILE is written as a sequential file by Subroutine NSECHO to store field annexation information row by row during the processing. BUFFER FILE is read by RDWRTE during the writing of the intermediate tape. The second file is also written by NSECHO to store the field statistics. This file is used by RDWRTE to compute the data values on the Intermediate Results Tape if the 'MAP' options has been selected (see control card description for NSLECHO).

BUFFER FILE

This file contains one record of length m bytes for every line processed by NSECHO. The length m equals four times the number of cells/line. The format of the record is I*2 with the following structure.

BUFFER(2i+1) = -1 if cell i is not homogeneous.
 The relative field number of cell i if cell i is homogeneous.

BUFFER(2i+2) = -1 if cell i is not homogeneous.
 Absolute statistics record number (in the STAT SCRATCH
 disk file) if cell i is homogeneous.

STAT SCRATCH

This file contains one record of $2+NC+NC*(NC+1)/2$ words for every field opened by NSECHO where NC is the number of channels considered by NSECHO. Each record has the following structure:

STAT(1)	I*2 Relative field number.
STAT(2)	I*2 Absolute sequence number of the field.
STAT(3)	I*4 Number of pixels in the field used to compute these statistics.
STAT(3+1)	R*4 Means for i^{th} channel if field STAT(2).
STAT(3+NC+1) until STAT(3+NC+NC*(NC+1)/2)	R*4 Correlation matrix of the field.

The BUFFER FILE is written sequentially using unformatted fortran IO. STAT SCRATCH is written using system support routine DEFINE FILE and unformatted fortran IO.

INTERMEDIATE TAPE FILE

This file is written on tape by Subroutine RDWRTE in the Nonsupervised ECHO processor's Field Extraction Phase and becomes the primary input to NS2ECHO (the Classification Phase). The Nonsupervised ECHO Intermediate Tape File is composed of six different types of data records. One each of record types 1, 2, and 3 are written on each Intermediate Tape File, one record type 4 is written for each set of field statistics recorded in the STAT SCRATCH file (one covariance matrix and a vector of channel-means for each field isolated by NSECHO), and one type 5 followed by a type 6 for every line processed in the Nonsupervised ECHO Field Extraction Phase.

Record Type 1 (ID Record)

This is similar to the conventional 800 Byte LARSYS ID record described in the LARSYS SYSTEM MANUAL 1. This records structure is as follows:

<u>Bytes</u>	<u>Format</u>	<u>Size</u>	<u>Description</u>
1-4	I*4	1 word	Intermediate tape number
5-8	I*4	1 word	Intermediate tape file number
9-12	I*4	1 word	Intermediate run number
13-16	I*4	1 word	Number of fields isolated by the field extraction phase
17-20	I*4	1 word	Number of data channels
21-24	I*4	1 word	Number of data samples per channel per line
25-40	Alpha	4 words	Flightline Identification (16 characters)
41-44	I*4	1 word	Month data was taken
45-48	I*4	1 word	Day data was taken
49-52	I*4	1 word	Year data was taken
53-56	Alpha	1 word	Time data was taken
57-60	I*4	1 word	Altitude of aircraft
61-64	I*4	1 word	Ground heading of aircraft
65-76	Alpha	2 words	Date data run was generated (12 characters)
77-80	I*4	1 word	Number of lines in this run

<u>Bytes</u>	<u>Format</u>	<u>Size</u>	<u>Description</u>
81-82	I*2	½ word	Number of the first channel used by NSECHO
83-84	I*2	½ word	Calibration code of the first channel used by NSECHO
85-200	I*2		For each channel used repeat information of the half words in bytes 81-82 and 83-84. The remaining bytes are equal to 0.
201-203	R*4	1 word	Lower limit in Micrometers of the first spectral band on the <u>original</u> MIST tape
205-208	R*4	1 word	Upper limit in Micrometers of the first spectral band on the <u>original</u> MIST tape
209-212	R*4	1 word	The suggested value of "C0" calibration pulse for the first spectral band
213-216	R*4	1 word	The suggested value of "C1" calibration pulse for the first spectral band
217-220	R*4	1 word	The suggested value of "C2" calibration pulse for the first spectral band
221-800	R*3		Repeat of words in bytes 200 to 220 for the channels on the <u>original</u> MIST tape. The remaining bytes are set to 0.

Record Type 2

This record is 17 fullwords long. It describes the area which has been processed by the Field Extraction Phase of the Nonsupervised ECHO processor to produce the Intermediate Tape File.

<u>Bytes</u>	<u>Format</u>	<u>Description</u>
1-4	I*4	MIST run number of processed area
5-12	Alpha	Field designation on field description card
13-16	I*4	Beginning line number
17-20	I*4	Last line number
21-24	I*4	Line interval
25-28	I*4	First column number

<u>Bytes</u>	<u>Format</u>	<u>Description</u>
29-32	I*4	Last column number
33-36	I*4	Column interval
37-68	Alpha	Information from columns 51-80 on the field description card

Record Type 3

This record is 33 words long and stores the parameters used in the Nonsupervised Field Extraction Phase.

<u>Bytes</u>	<u>Format</u>	<u>Parameters</u>
1-4	R*4	Cell width
5-8	R*4	Variance test threshold
9-12	R*4	Mean test threshold
13-16	R*4	Homogeneity test threshold for the first channel used
17-132	R*4	Homogeneity test thresholds for the remaining channels

Record Type 4

This record is the same as on the STAT SCRATCH file. There are m records of type 4, where m is the number of fields isolated by the Field Extraction Phase of the Nonsupervised ECHO processor. Each record has $2+NC+NC*(NC+1)/2$ words where NC is the number of channels used by NSECHO. The structure of each record is:

<u>Size</u>	<u>Format</u>	<u>Contents</u>
1-2	I*2	Relative field number
3-4	I*2	Absolute sequence number of the field
5-8	I*4	Number of pixels in the field with this statistics
9-(NC+2)*4	R*4	Mean value for the 1 st , 2 nd , . . . NC th channel used by NSECHO
(NC+2)*4+1 - (NC+2)*4+NC*(NC+1)*2	R*4	Correlation matrix of the field

Record Type 5

This is similar to a standard data line in a LARSYS MIST tape. If the MAP option was specified, the original data has been altered so that for the pixels identified as falling in an object, the channel mean of the object replaces the raw data value. If the MAP option is not active, the raw data is copied unaltered to the intermediate tape.

Each data record will contain one scan line of data from ID(5) (see ID Record) channels. The first halfword (2 bytes) of the record will be the line number. The second halfword (2 bytes) will be the roll parameter (which is a number indicating relative position of the roll of the aircraft for this line of data). If the roll parameter is -32,767, the data for the given line does not exist. If the roll parameter has not been calculated, it will be set to 32,767. The fifth byte will be the first sample from the requested channel. The sixth byte will be the second sample from the first requested channel, and so on through ID(6) samples and ID(5) channels. A Type 5 record will be $ID(5)*ID(6)+4$ bytes long.

All data for each channel is from the field of view of the scanner except the last six bytes. The last six are calibration data in the order of appearance.

1. C_0 "0" or dark level
2. VC_0 Variance of C_0
3. C_1 Calibration source C_1
4. VC_1 Variance of C_1

5. C_2 Calibration source C_2

6. VC_2 Variance of C_2

where C_i - Calibration value i and VC_i - calculated variance of calibration value i

During the reformatting process a record may be had due to tape or other errors. When this happens, the data roll parameter and calibration points will be set to zero. On good data records all data and calibration values will be in the range of 0 to 255 (bit form) with no sign included in the eight bits. A data value of 0 to 255 means that the data point was cut off during the digitization process. Data values then range between 0 and 255 with 0 indicating low relative irradiance and 255 indicating high relative irradiance.

Record Type 6

Identical to a record in BUFFER FILE, i.e. it has m words for each line, processed, where m is the number of cells per line. The structure of each record is:

<u>Bytes</u>	<u>Format</u>	<u>Contents</u>
1-2	I*2	-1 if the cell is singular, otherwise relative field number of the cell
3-4	I*2	-1 if the cell is singular, otherwise number of the statistics record of the field the cell belongs to
5-2m	I*2	Similar to the first two halfwords describing the nature of the remaining ($m-1$) cells

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